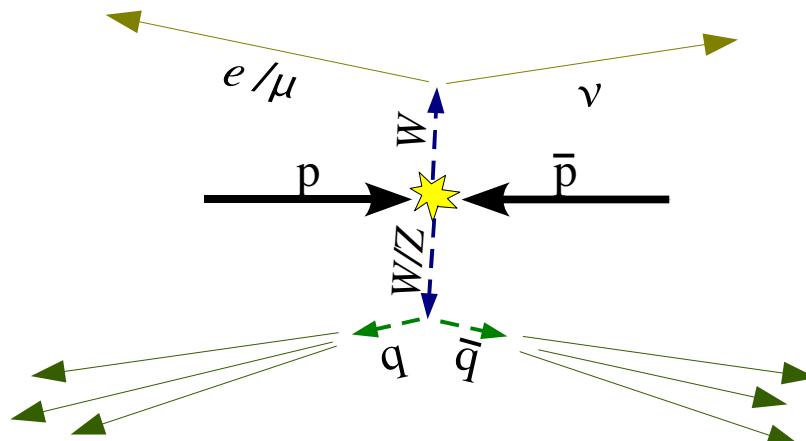




First Evidence for WW and WZ Diboson Production with Semi-leptonic Decays at a Hadron Collider

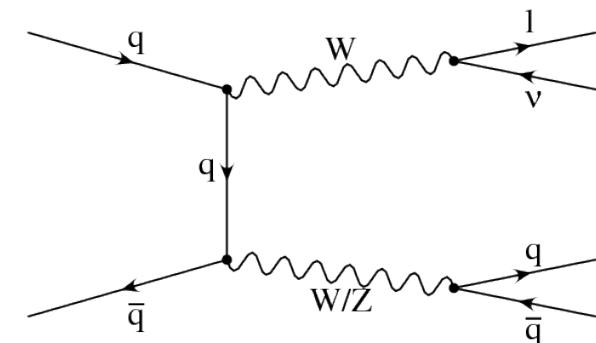
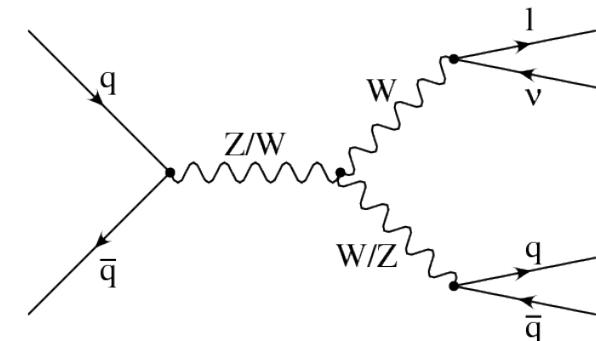


Joseph G. Haley
(D0 Collaboration)

April 14, 2009

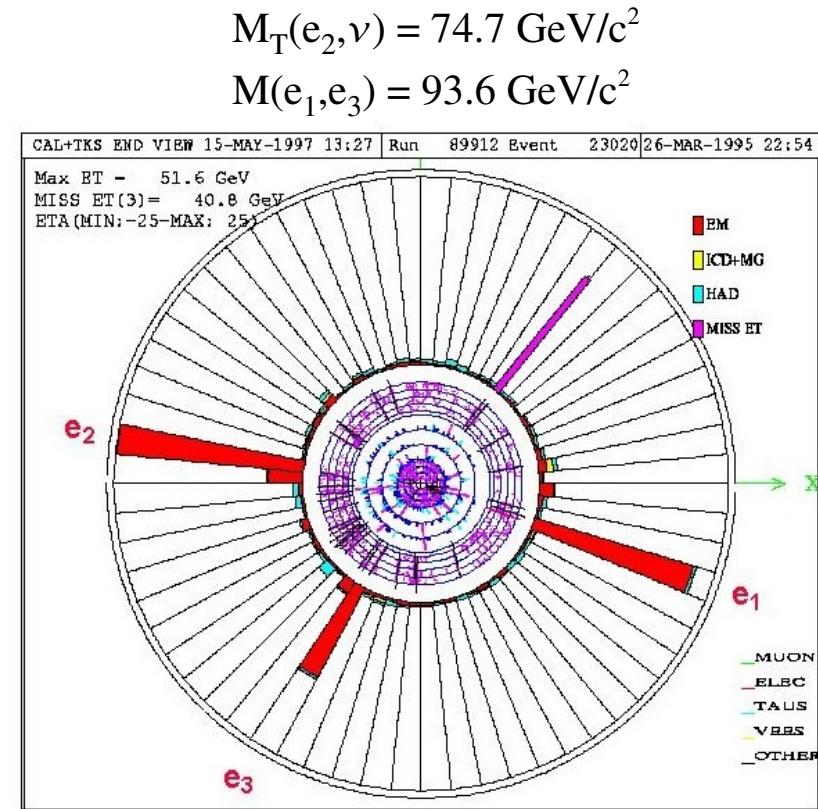
Outline

- Motivation for Studying $WW/WZ \rightarrow l\nu qq$
- The Apparatus
 - Accelerator Chain, D0 Detector
- Samples and Event Selection
 - Data, SM Simulation
- MC Corrections and Systematic Uncertainties
 - Efficiencies, Energies, MC Modeling
- Multivariate Discrimination
- The Measurement
 - Cross Section, Significance
- Summary
- Continuing Work



Motivation: Electroweak Studies

- Probe of the Electroweak sector of the SM at the Tevatron
- Deviations in production rate or event kinematics would indicate new physics
- Currently all diboson measurement are in fully leptonic final states
 - $WW \rightarrow l\nu l\nu$
 - DZero $\sim 1.0/\text{fb}$: $\sigma(WW) = 12.3 \pm 2.0 \text{ pb}$
 - CDF $\sim 0.83/\text{fb}$: $\sigma(WW) = 13.6 \pm 3.1 \text{ pb}$
 - SM prediction: $\sigma(WW) = 12.4 \pm 0.8 \text{ pb}$
 - $WZ \rightarrow l\nu ll$
 - DZero $\sim 1.0/\text{fb}$: $\sigma(WZ) = 2.7^{+1.7}_{-1.3} \text{ pb}$
 - CDF $\sim 1.9/\text{fb}$: $\sigma(WZ) = 4.3^{+1.1}_{-1.1} \text{ pb}$
 - SM prediction: $\sigma(WZ) = 3.7 \pm 0.3 \text{ pb}$

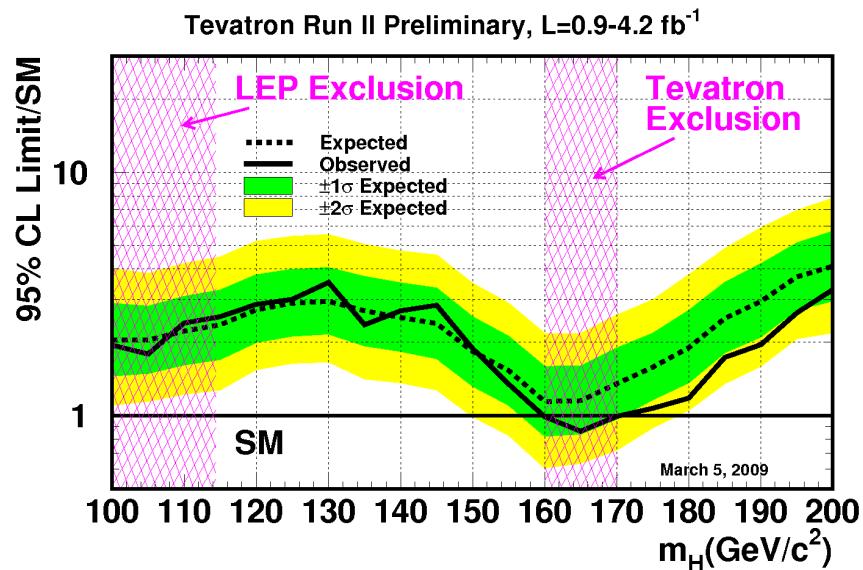
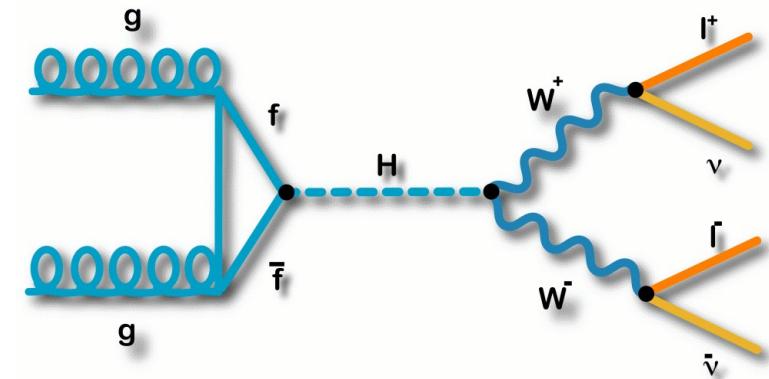
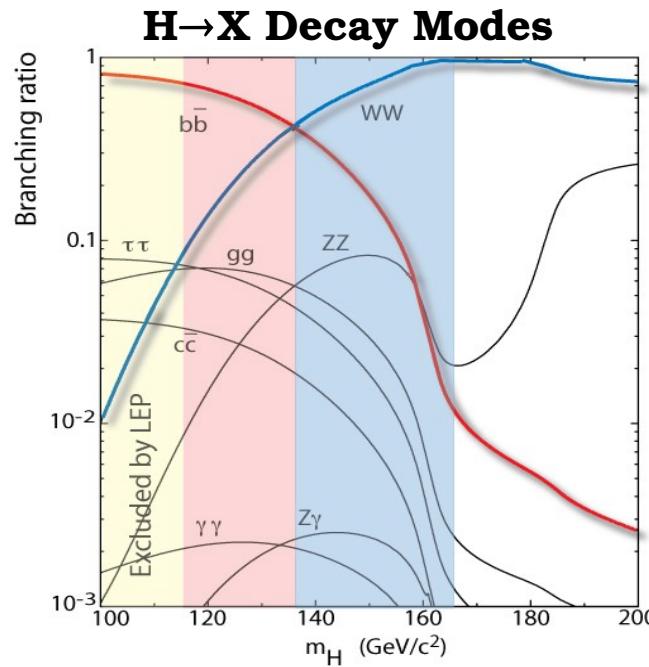


⇒ $WW/WZ \rightarrow l\nu qq$ complimentary to fully leptonic measurements



Motivation: Higgs Searches

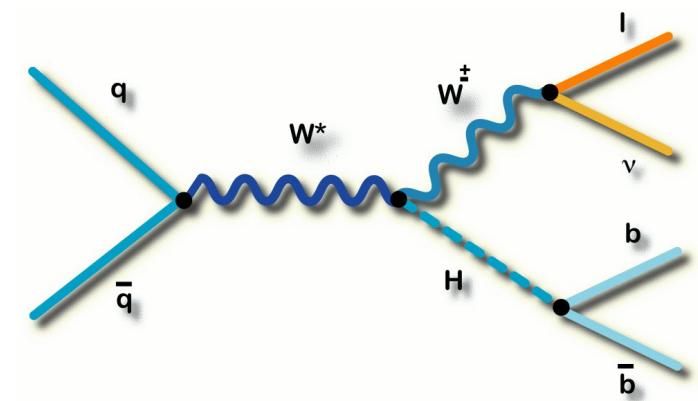
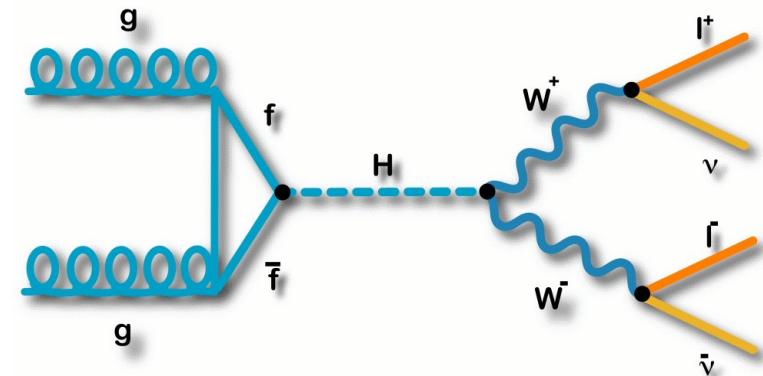
- $H \rightarrow WW$ is the dominant decay mode for a high mass Higgs ($m_H > 135 \text{ GeV}/c^2$)
 - Drives current exclusion limits
 - Direct diboson production is the single most important background
 - Important to understand this contribution



Motivation: Higgs Searches

- $H \rightarrow WW$ is the dominant decay mode for a high mass Higgs ($m_H > 135 \text{ GeV}/c^2$)
 - Drives current exclusion limits
 - Direct diboson production is the single most important background
 - Important to understand this contribution
- $WH \rightarrow l\nu bb$ is a promising search mode for low mass Higgs ($m_H < 135 \text{ GeV}/c^2$)
 - Similar topology/final state to $WW/WZ \rightarrow l\nu qq$
 - Similar challenges
 - ▶ Small signal in a large background:
 $WW/WZ \rightarrow l\nu qq$ B/S ≈ 35 , $WH \rightarrow l\nu bb$ B/S $\gtrsim 90$
 - Similar analysis techniques
 - ▶ Multivariate classifiers, statistical treatment

⇒ $WW/WZ \rightarrow l\nu qq$ is a proving ground for Higgs search



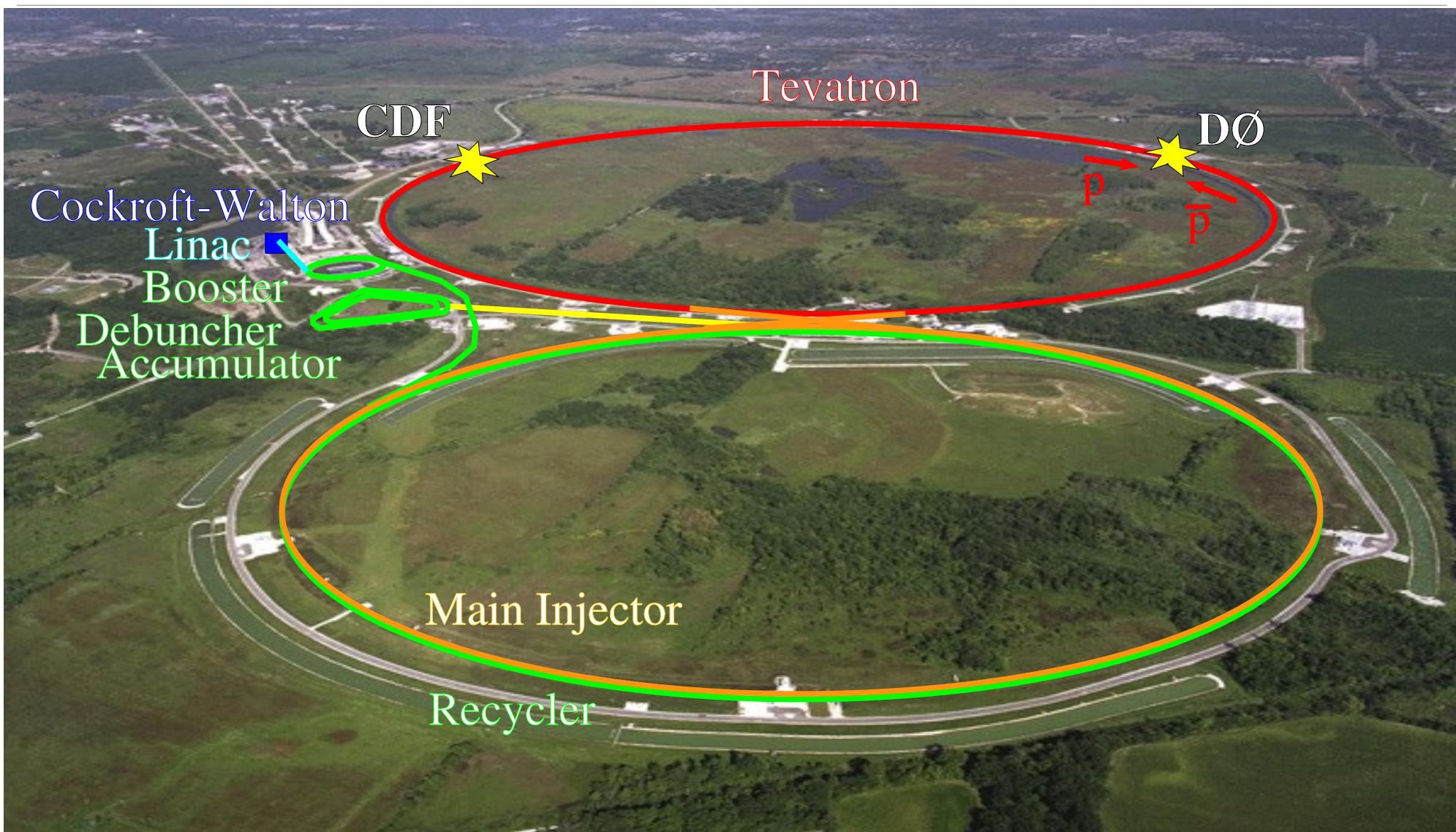


Apparatus

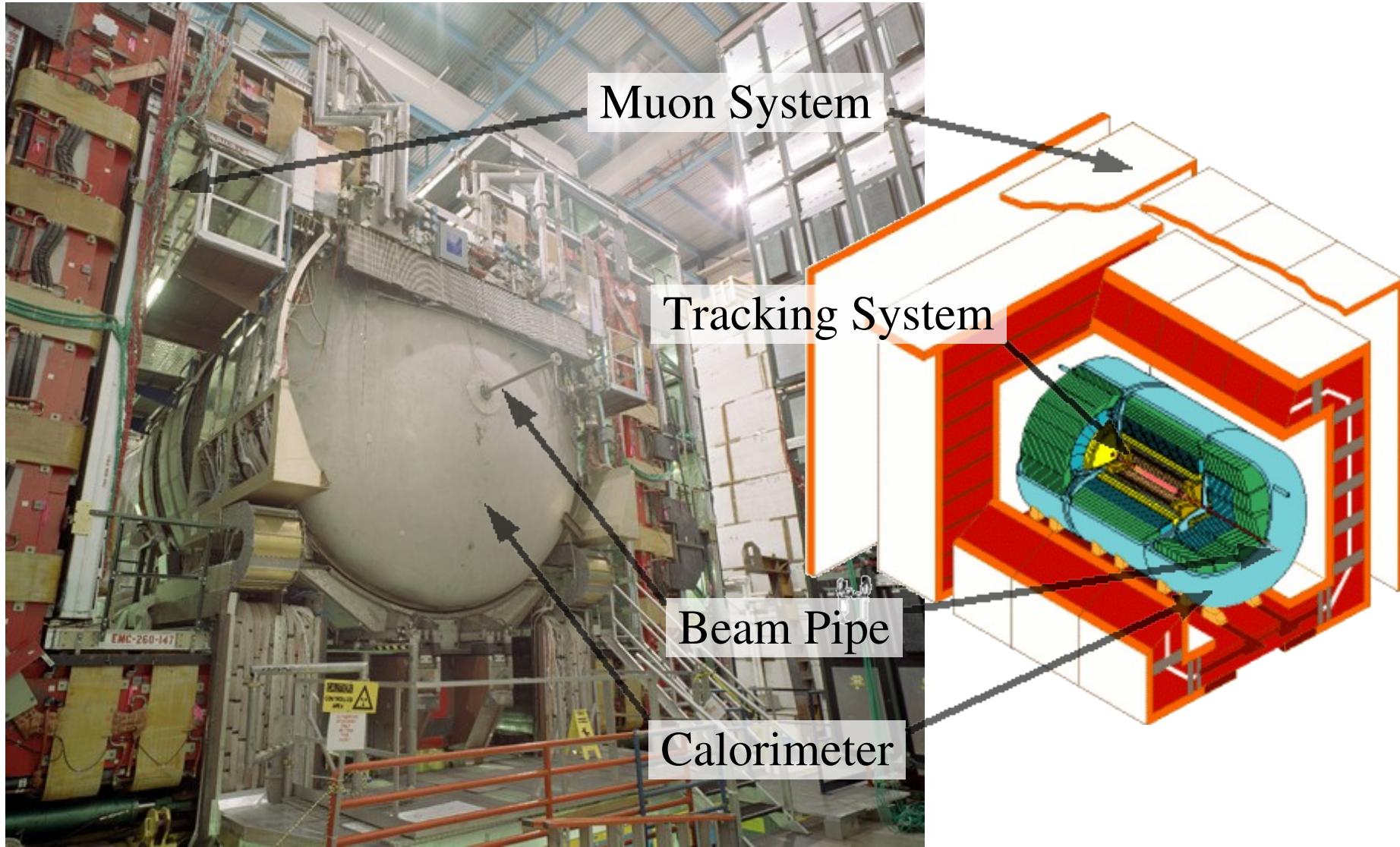




The Tevatron

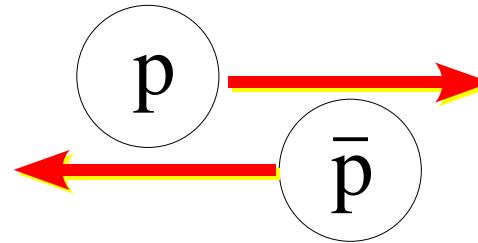


The DO Detector



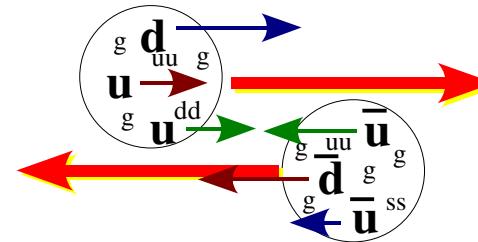
Hadron Collisions and Coordinate System

- Momentum of colliding protons and antiprotons divided among partons
 - Detector is not the zero-momentum-frame



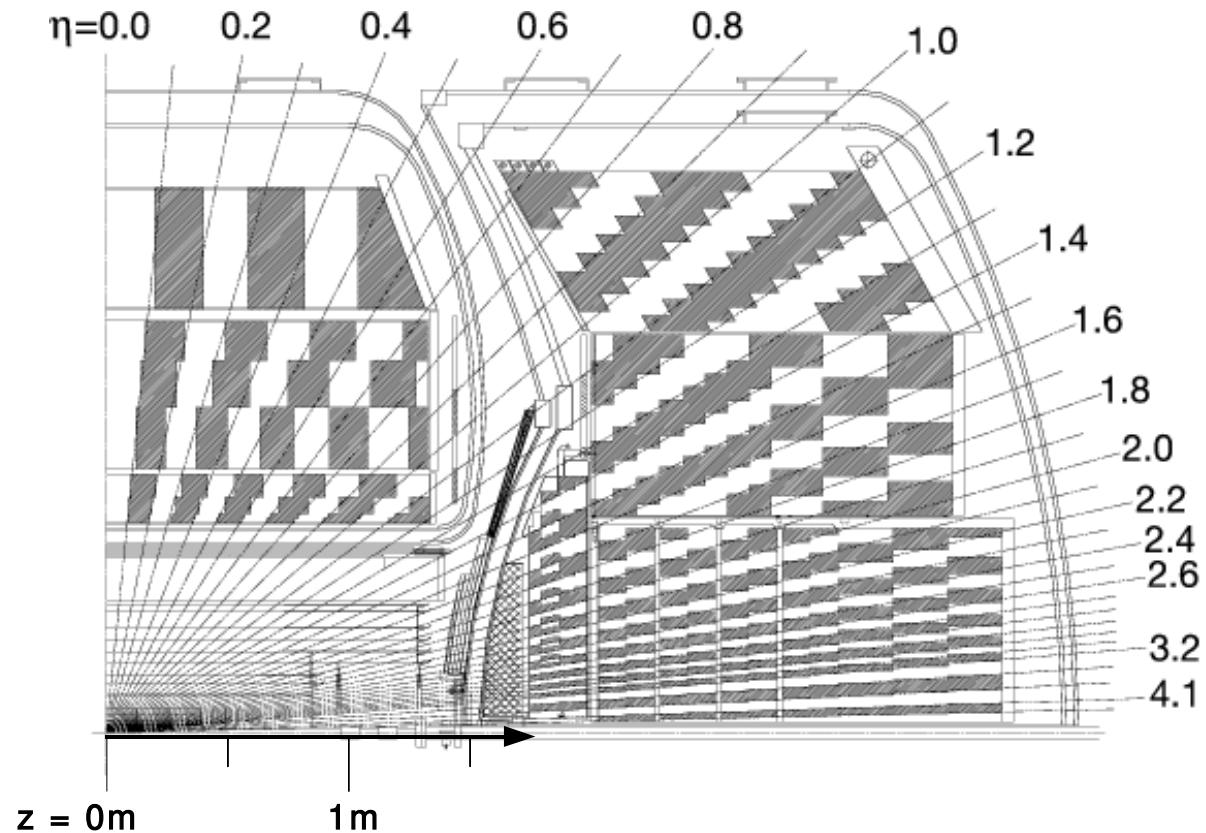
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 - $p_x = 0, p_y = 0$
 - Total longitudinal momentum is unknown
 - $p_z = ?$

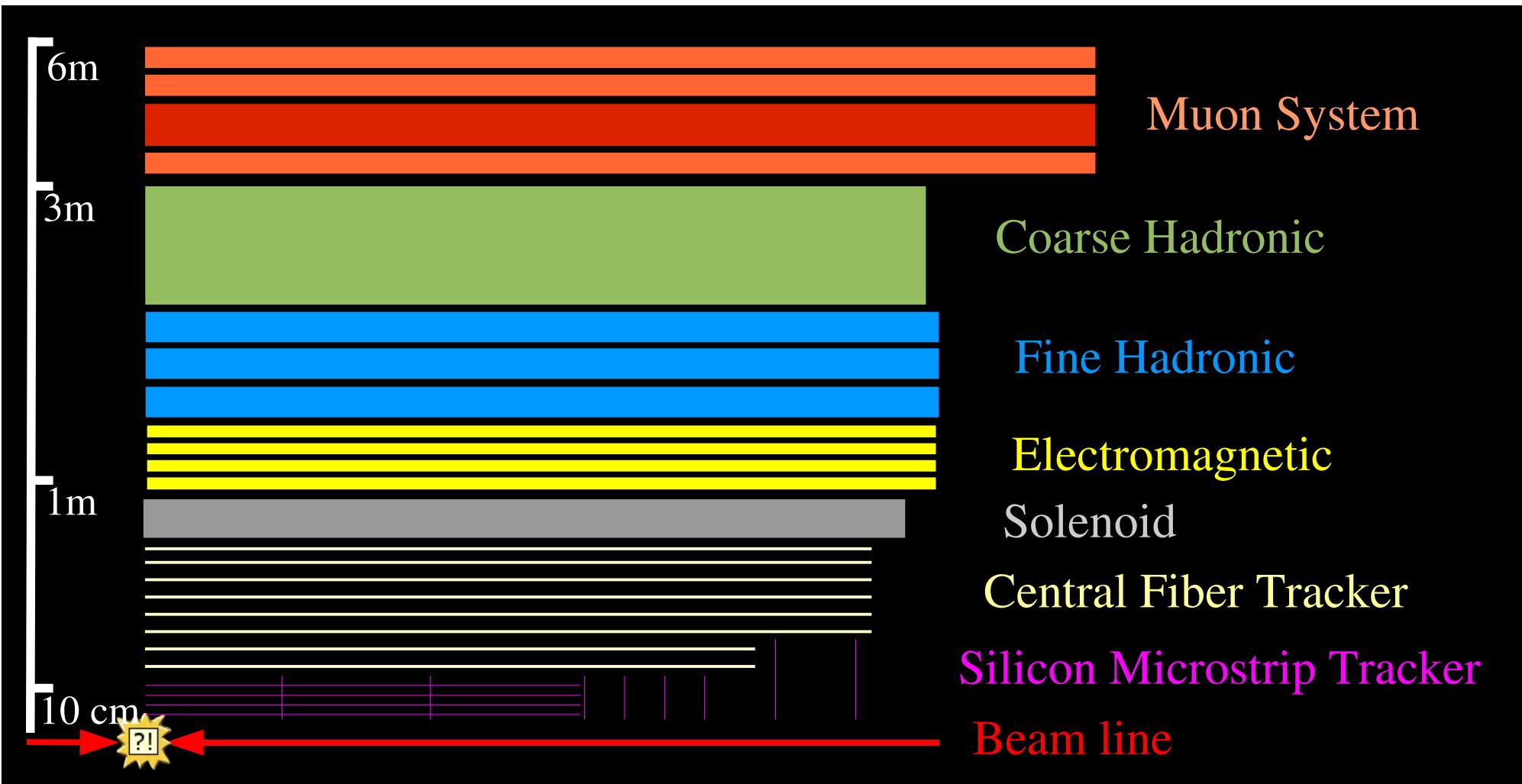


Hadron Collisions and Coordinate System

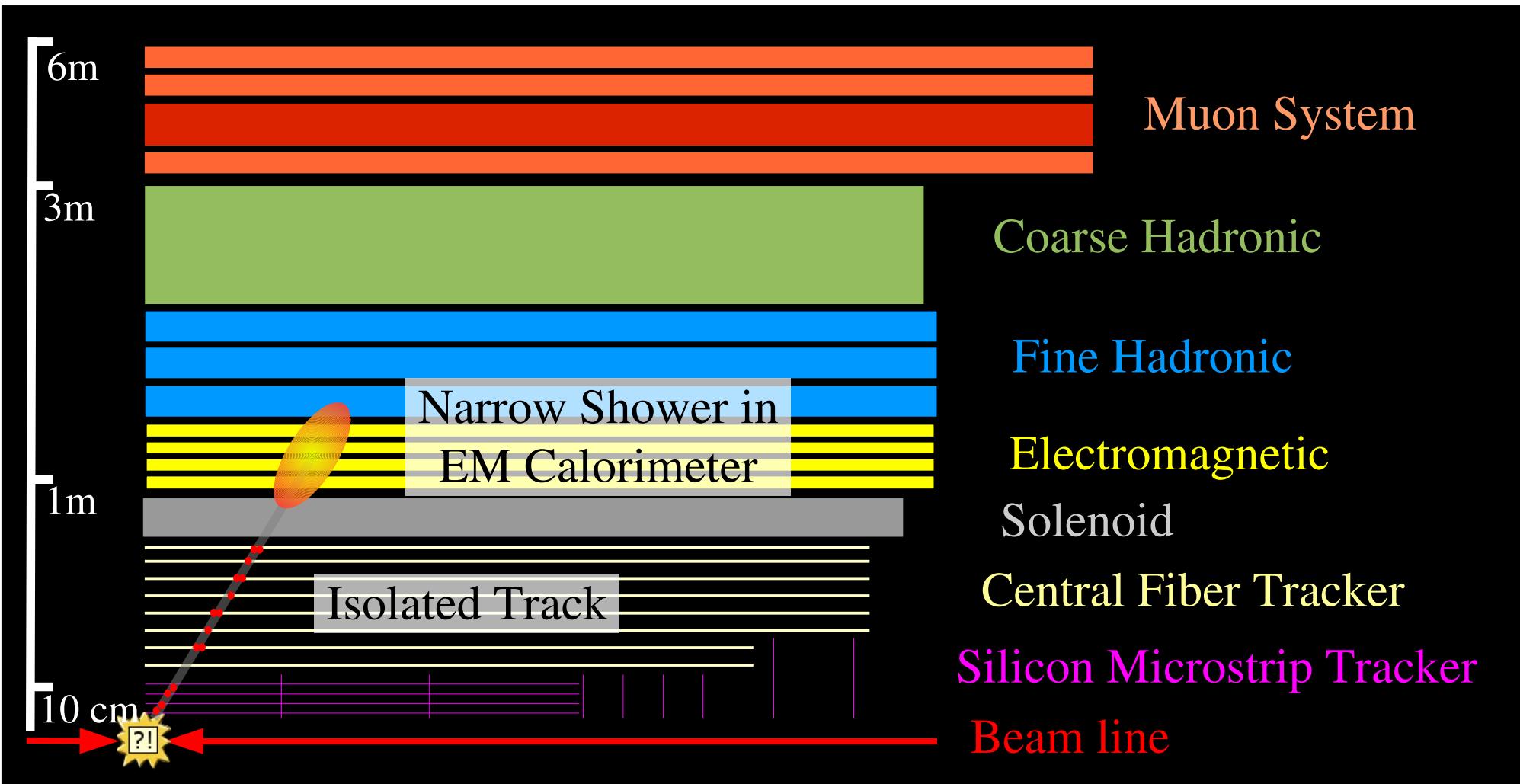
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 - Detector is not the zero-momentum-frame
 - Total transverse momentum is zero
 - $p_x = 0, p_y = 0$
 - Total longitudinal momentum is unknown
 - $p_z = ?$
 - Convenient to use pseudorapidity η instead of polar angle θ
 - $\eta \equiv -\ln[\tan(\theta/2)]$
 - Approaches Lorentz invariant rapidity for $E \gg m$



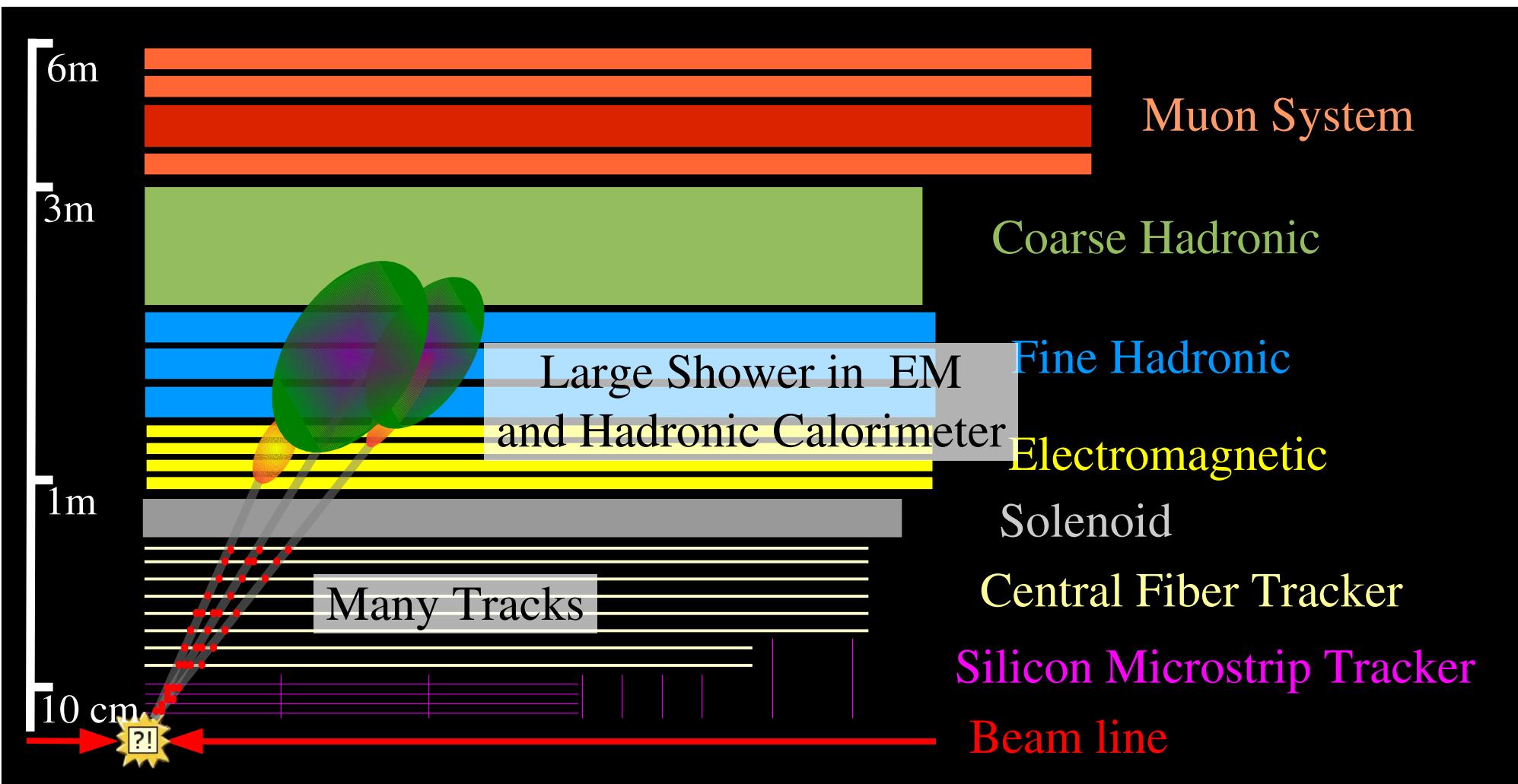
The $\mathcal{D}0$ Detector: A Schematic View



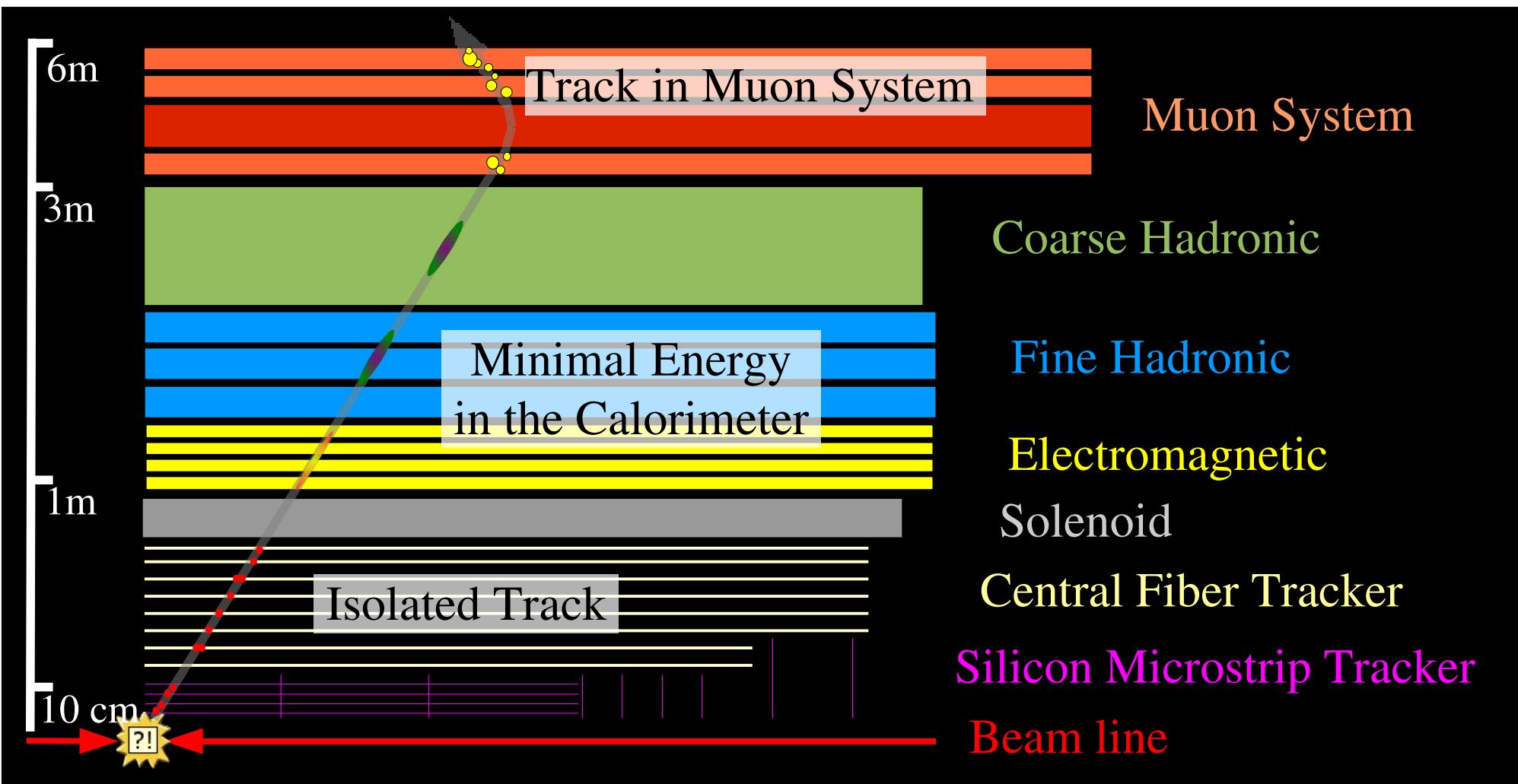
Electrons



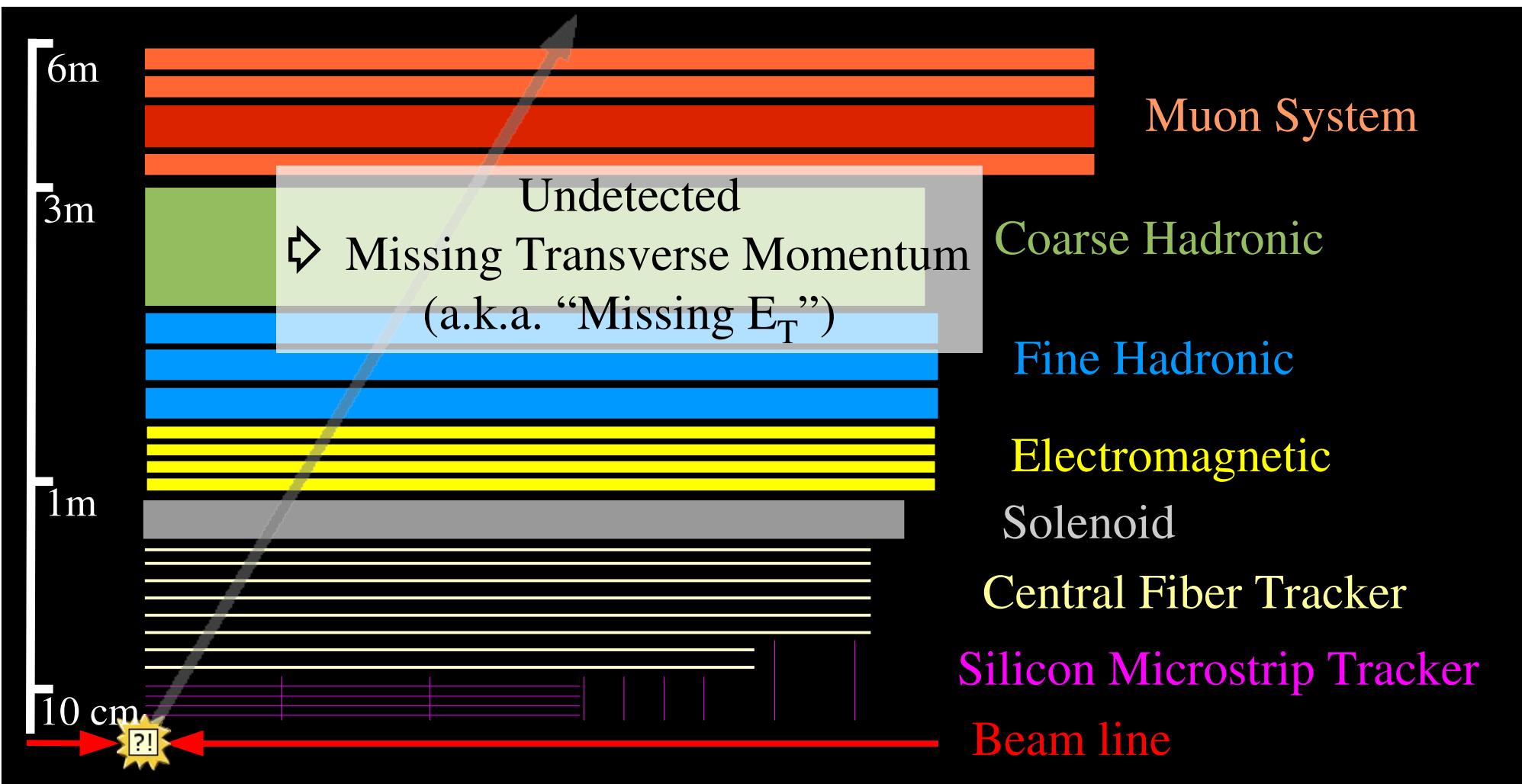
Jets



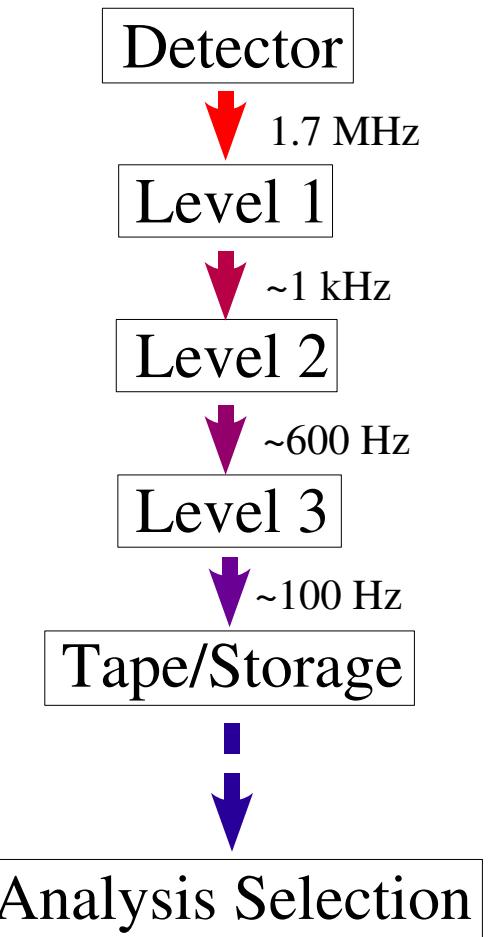
Muons



Neutrinos



Trigger and DAQ



My Contributions to the Trigger

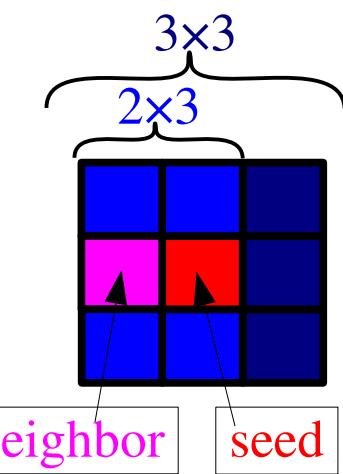
- Level 2 EM trigger upgrade for Run IIb
 - Needed to improve rejection to keep up with increased luminosity and improved Level 1 trigger
 - Constructed a Level 2 EM likelihood variable
 - Four inputs:

$$\frac{\text{EM } E_T \text{ seed}}{\text{TOT } E_T \text{ seed}}$$

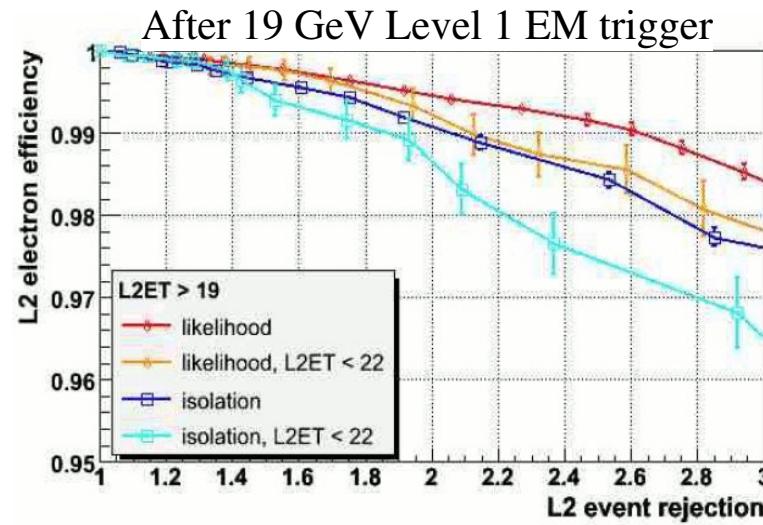
$$\frac{\text{EM } E_T \text{ neighbor}}{\text{EM } E_T \text{ seed}}$$

$$\frac{\text{EM } E_T \text{ } 3\times3}{\text{TOT } E_T \text{ } 3\times3}$$

$$\frac{\text{EM } E_T \text{ seed}}{\text{EM } E_T \text{ } 3\times3}$$



- Factor of ~3 background rejection with signal efficiency near 99%
- In some case, ~30% more rejection than Level 2 isolation variable

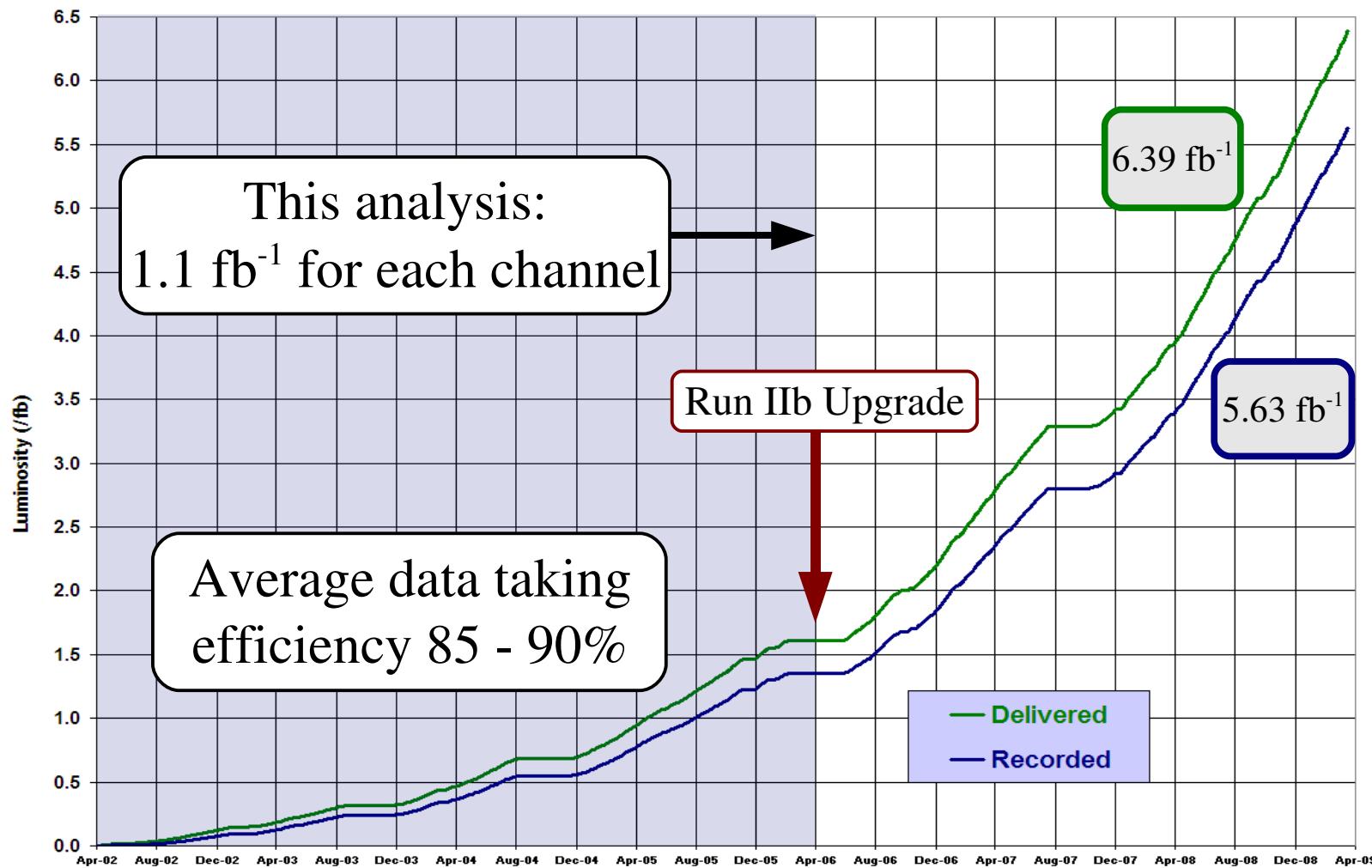


*Samples and
Event Selection*



Data Sample

- Dataset: Run IIa of the Tevatron (2002 – 2006)



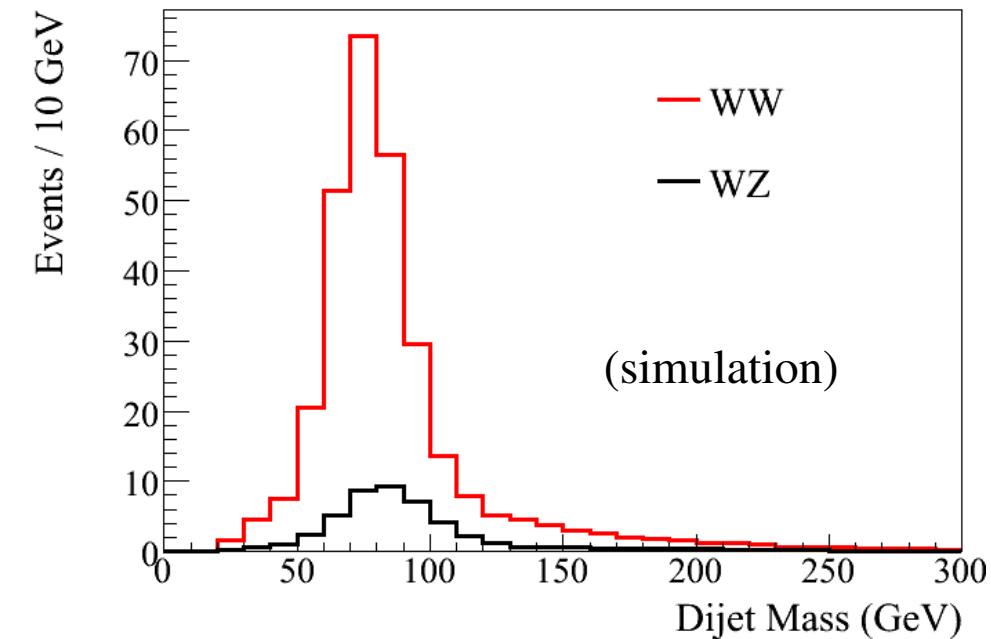
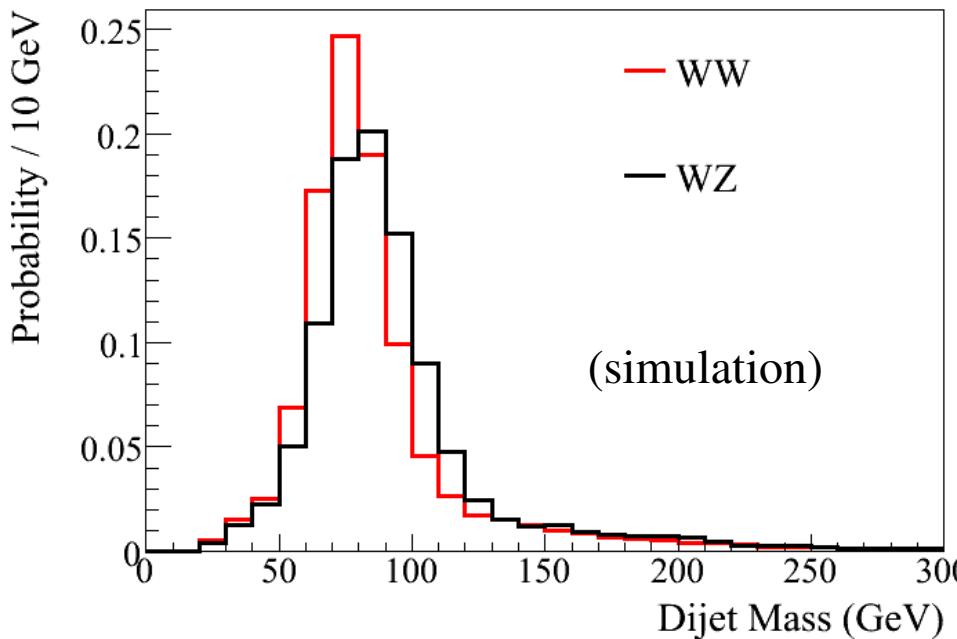
SM Expectation

- Simulated Samples
 - Monte Carlo generators simulate the SM prediction for signal and backgrounds with real high p_T leptons
 - ALPGEN+Pythia: $W+\text{jet}$, $Z+\text{jets}$, $t\bar{t}+\text{jets}$
 - COMPHEP+Pythia: Single-top
 - Pythia: Diboson Signal
- Multijet Estimation
 - Events without a real high p_T lepton
 - Probability for a jet to mimic a lepton is very small, but QCD processes have very large cross sections
 - Estimated from data using orthogonal selection
 - Corrected for contributions from samples modeled with MC

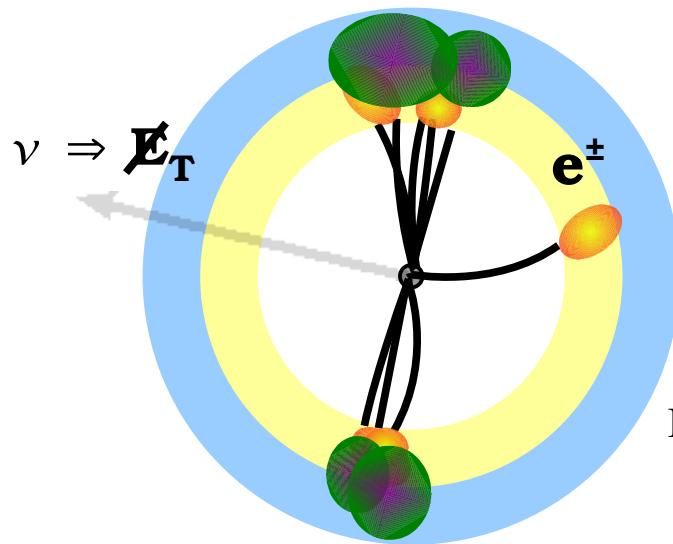


Note on WW vs. WZ

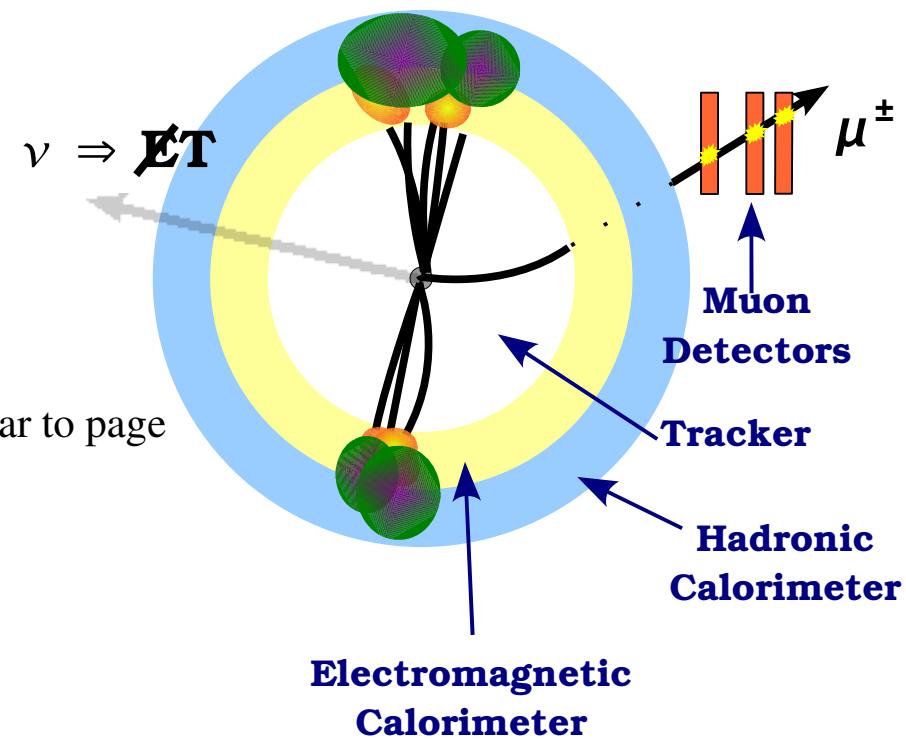
- WW and WZ treated as the same signal
 - Difference in mass of ~ 10 GeV/c 2
 - Dijet mass resolution ~ 20 GeV/c 2
- Relative contributions
 - $WW(WZ) \rightarrow l\nu qq \sigma \times BR \cong 3.5(0.5)$ pb



Event Selection



Beam perpendicular to page



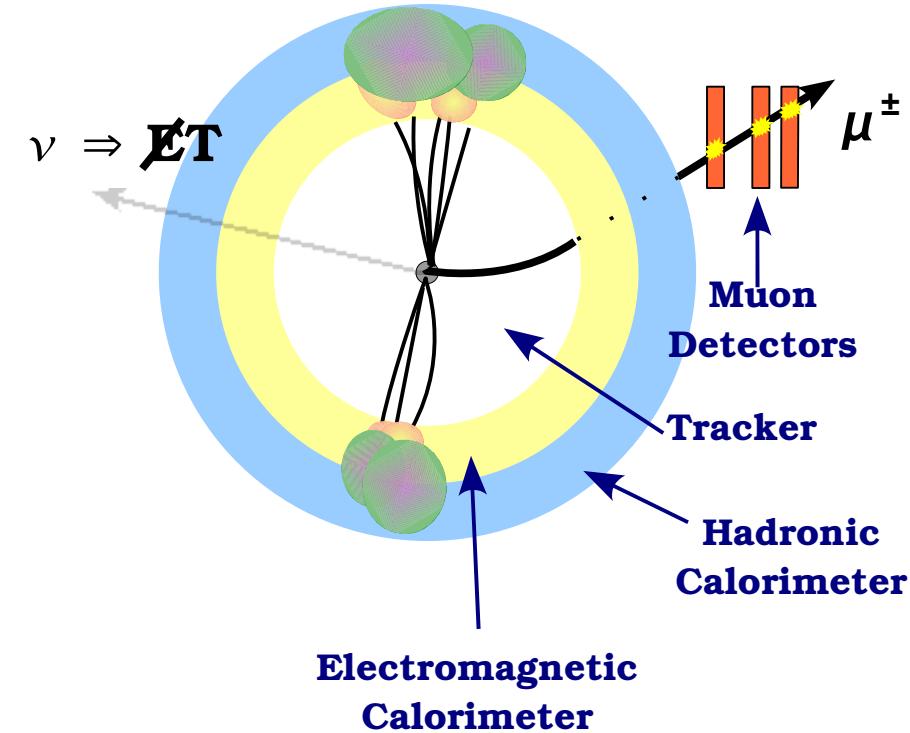
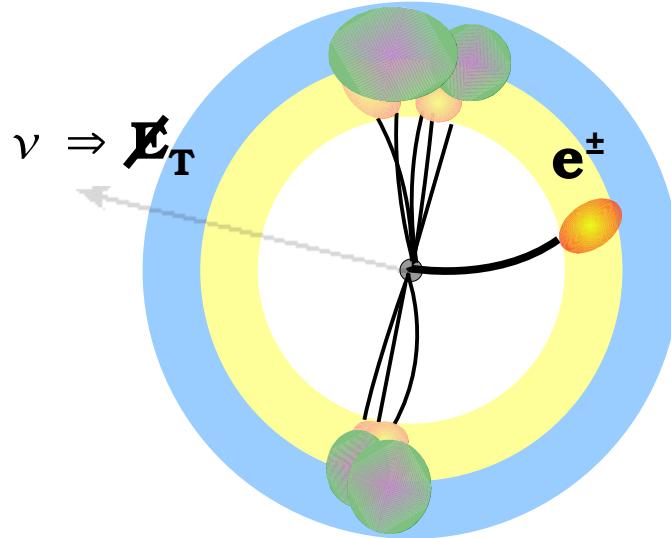
Event Selection

- $W \rightarrow l\nu$ Selection

- Electron:
 - $p_T \geq 20 \text{ GeV}/c$, $|\eta| < 1.1$
 - Isolated track and EM shower
 - Electron shower shape requirements

OR

- Muon:
 - $p_T \geq 20 \text{ GeV}/c$, $|\eta| < 2.0$
 - Hits in all three muons layers
 - Isolated in tracker and calorimeter



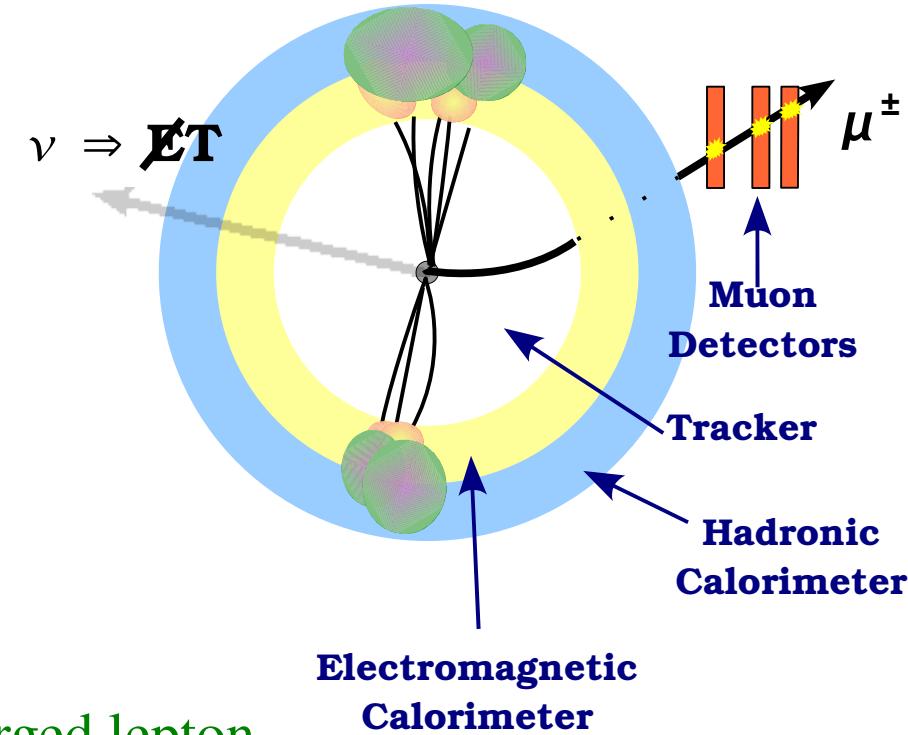
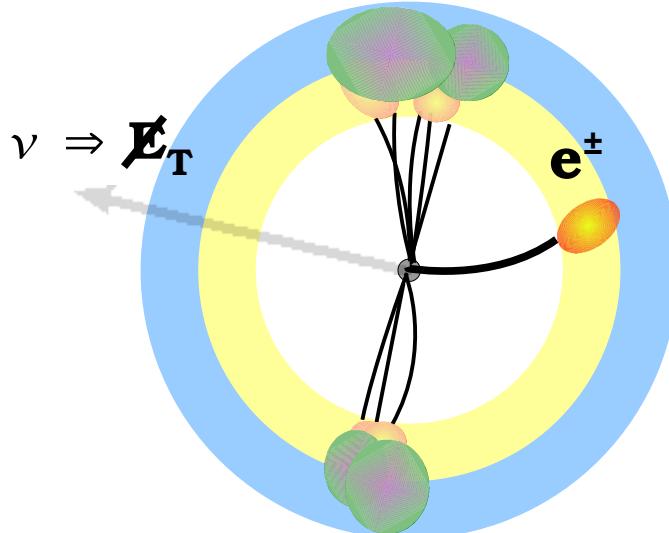
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OR

- Muon:
 - $p_T \geq 20 \text{ GeV}/c$, $|\eta| < 2.0$
 - Hits in all three muons layers
 - Isolated in tracker and calorimeter



- Neutrino \Rightarrow Missing $E_T \geq 20 \text{ GeV}$
- Transverse W mass $\geq 35 \text{ GeV}/c^2$
- Remove events with more than one charged lepton

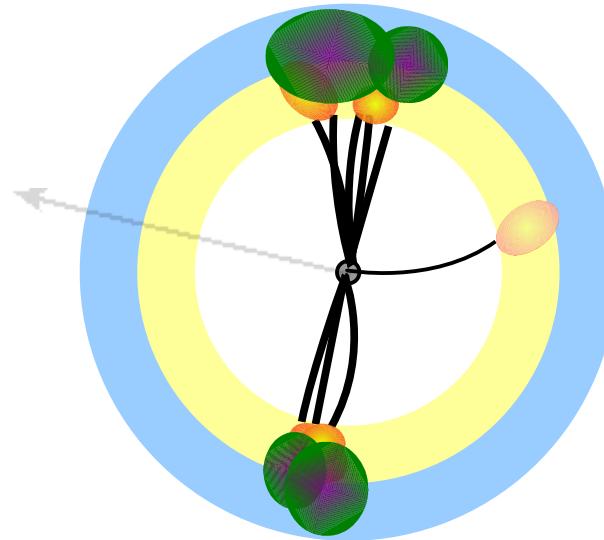


Event Selection

- $W/Z \rightarrow qq$ Selection

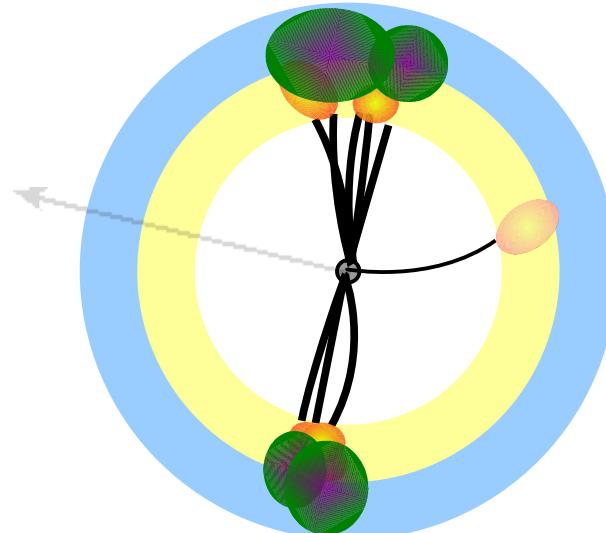
- At least two jets with radius $R = 0.5$ (cone algorithm)
- Leading jet: $p_T \geq 30 \text{ GeV}/c$, $|\eta| \leq 2.5$
- Second jet: $p_T \geq 20 \text{ GeV}/c$, $|\eta| \leq 2.5$

$$R \equiv \sqrt{(\Delta\phi + \Delta\eta)}$$



Event Selection

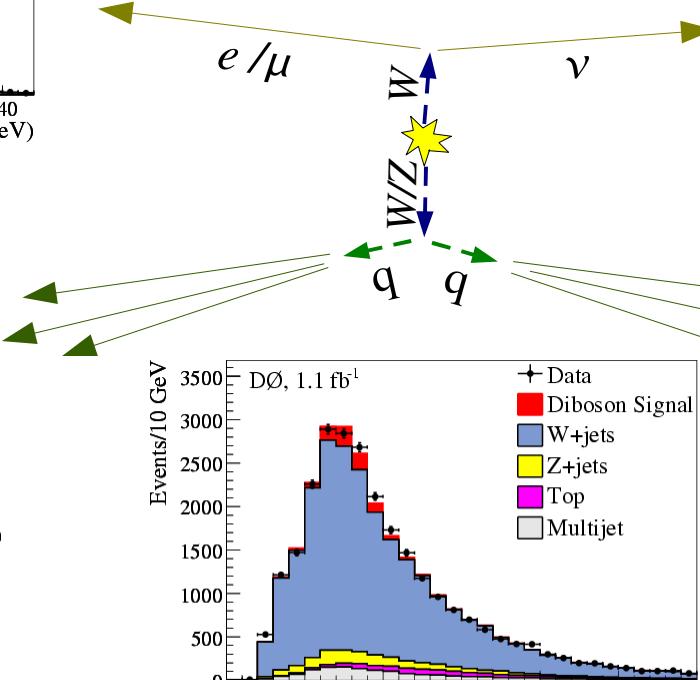
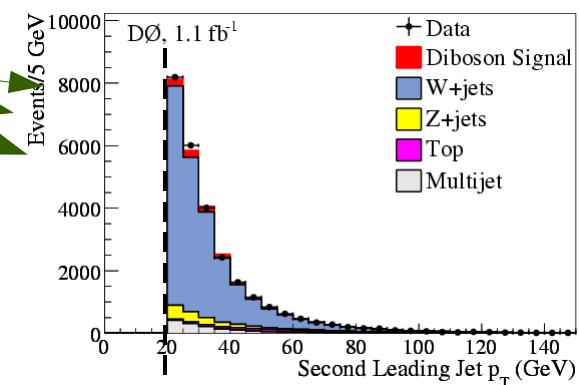
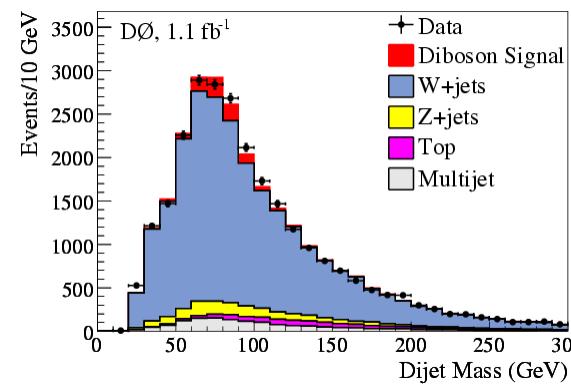
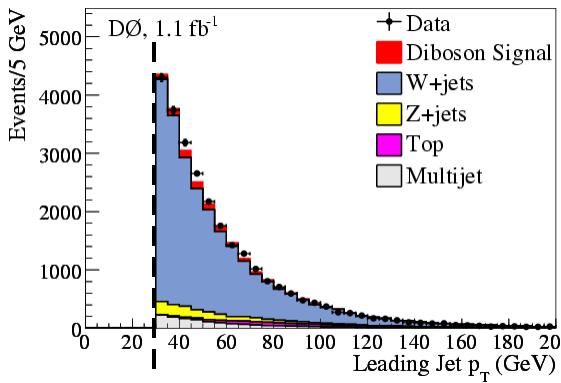
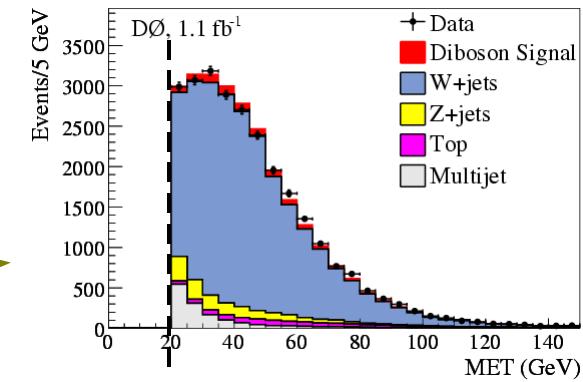
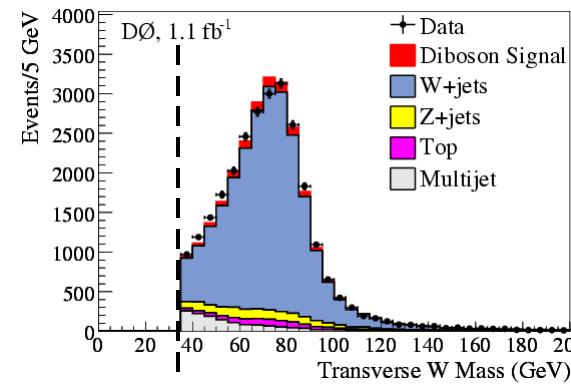
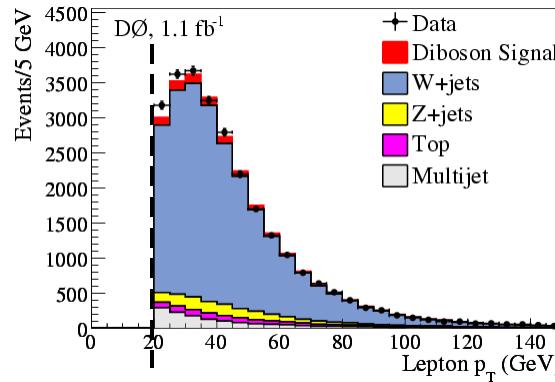
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Selected Events

Source	$e\nu qq$	$\mu\nu qq$
Diboson	360.5 ± 2.3	427.3 ± 2.7
$W + \text{jets}$	10226 ± 76	12012 ± 88
$Z + \text{jets}$	408 ± 13	1239 ± 20
Top	463.3 ± 2.2	437.0 ± 2.2
Multijet	825 ± 11	327.0 ± 9.6
Total Predicted	12283 ± 78	14442 ± 91
Data	12473 ± 112	14392 ± 120
$S/\sqrt{S+B}$	3.25	3.56

Event Selection

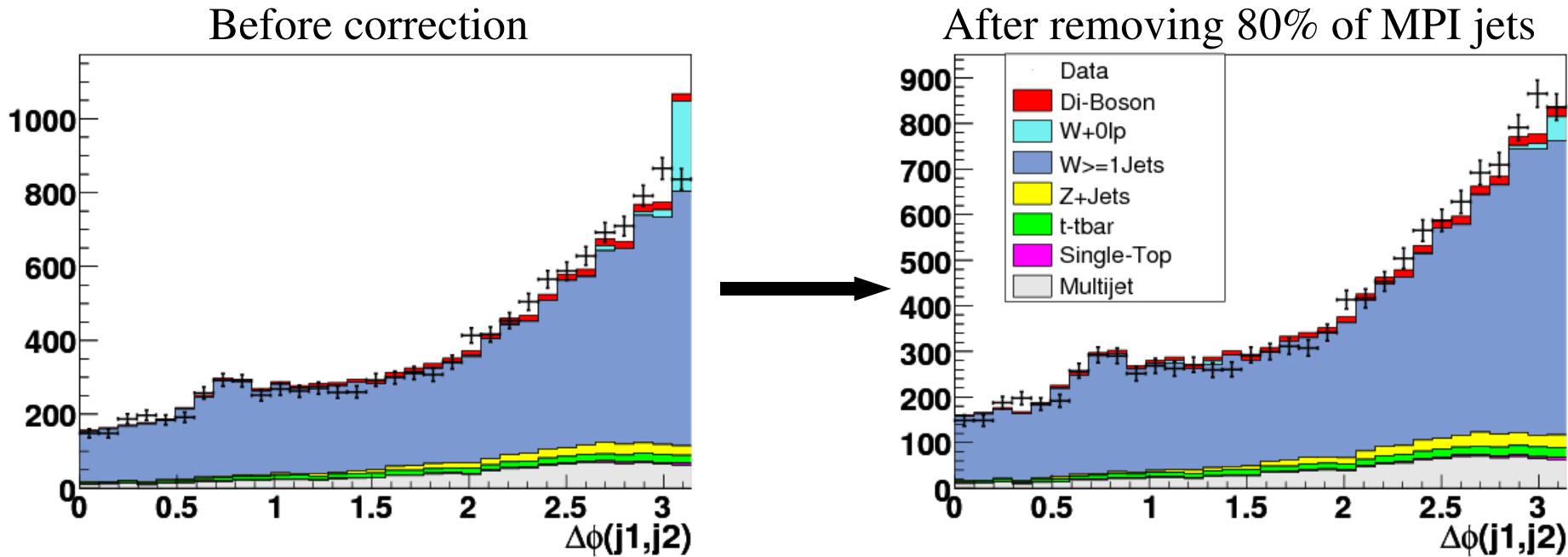


Monte Carlo Corrections and Systematic Uncertainties



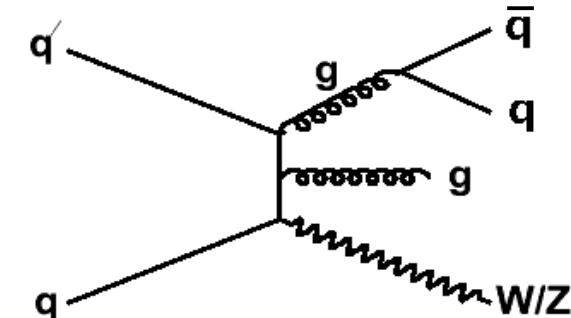
Modeling of Multiple Parton Interactions

- ALPGEN+Pythia MC generated with too many jets from multiple parton interactions (MPI)
 - Jets from more than one interaction in the proton-antiproton collision
 - Result in two jets that are back-to-back in ϕ
 - Mainly affects $W+0$ jet and $W+1$ jet events \Rightarrow too many pass the selection
 - χ^2 fit to data showed the number of selected MPI jets should be 80% less



ALPGEN W +jets Modeling

- Selected sample is dominated by W +jets background
 - W +jets is ~ 30 times larger than the signal!
 - Must accurately model W +jets at the percent level
- Potential problems with modeling of W +jets
 - Angles
 - Jet η and ΔR between jets is not modeled well by ALPGEN
 - To correct these differences we re-weight the MC to agree with data
 - Energies
 - ALPGEN does a fairly good job of modeling energies, however, energy distributions can be affected by ALPGEN generator and matching parameters
 - Studied the effects of these parameters to determine the uncertainty from these parameters
 - Determine best values for agreement with data
 - Cross section
 - The W +jets cross section is unconstrained when measuring diboson cross section



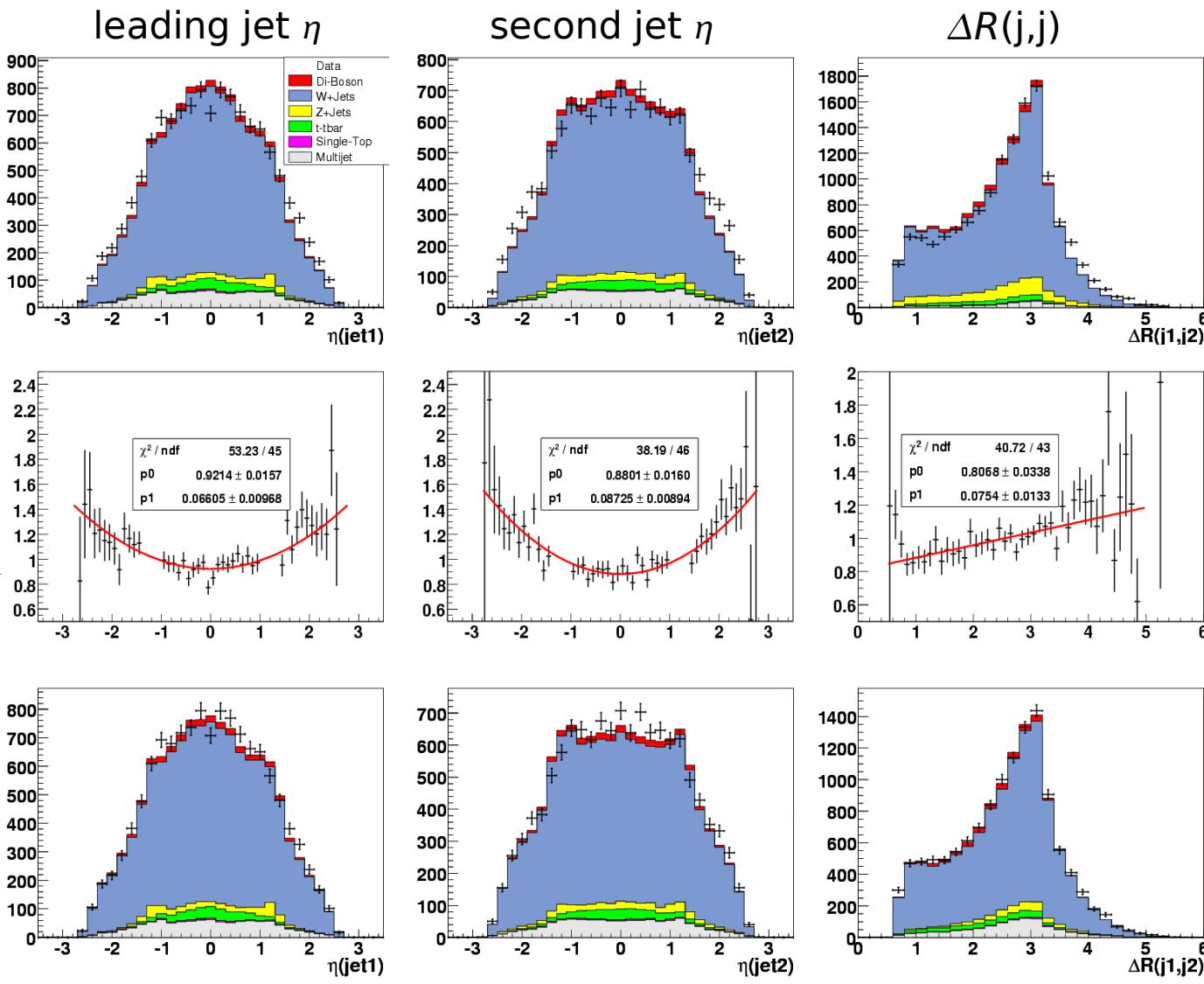
\mathcal{ALPGEN} Modeling: Angular Corrections

Poor agreement before:

Re-weight $W+jets$ MC by
a function fit to:

$$\frac{\text{Data} - (\text{non } W+jets \text{ MC})}{W+jets \text{ MC}}$$

(Include uncertainty on fit from
varying the signal cross section
from $0-2 \times$ the SM prediction)



Improved agreement after:





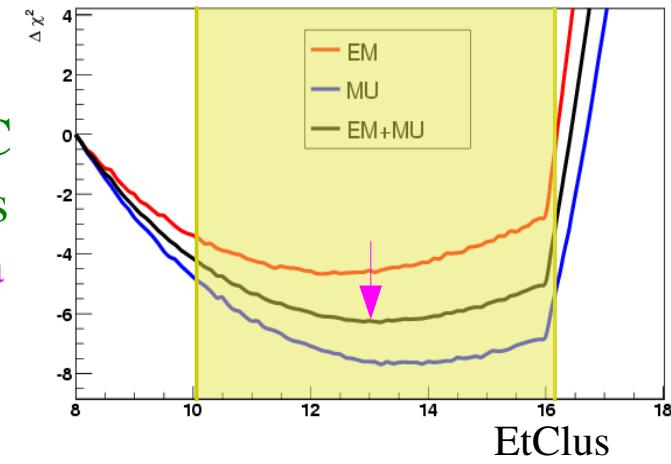
ALPGEN Modeling: Generator Parameters

- After correcting large angular discrepancies we can look at the smaller effects from ALPGEN generator and matching parameters
 - Qfac: Multiplies renormalization scale (Q-scale) by this factor
 - Ktfac: Multiplies parton kt scale (a.k.a. kperp or k_{\perp}) by this factor
 - RClus: Radius for clusters used in ALPGEN parton-jet multiplicity matching
 - **EtClus: E_T threshold for clusters used in parton-jet matching**
 - ▶ ALPGEN parton-jet matching required to keep from double counting events



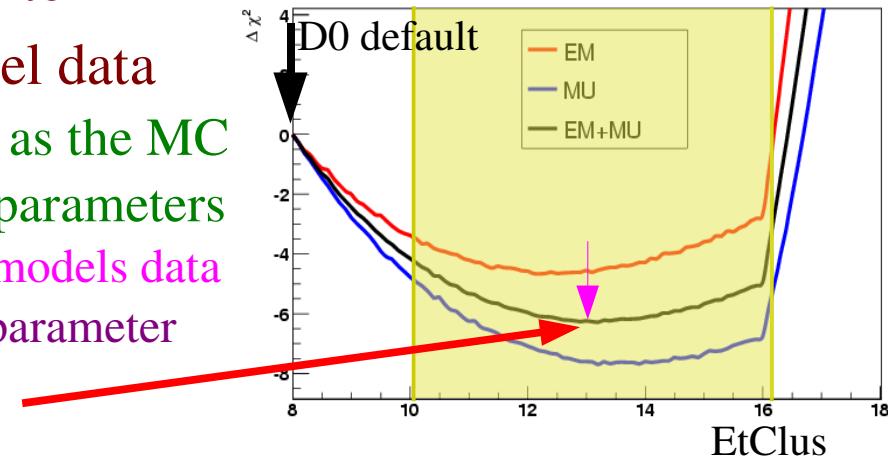
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 - ALPGEN parton-jet matching required to keep from double counting events
- Remove signal events ($55 \text{ GeV}/c^2 < m_{jj} < 110 \text{ GeV}/c^2$)
- Use p_T of leading jet and recoiling $W \rightarrow l\nu$ to determine parameter values that best model data
 - Measure change in χ^2 between data and MC as the MC shape is morphed by varying these ALPGEN parameters
 - Minimum χ^2 tells the parameter value that best models data
 - $\Delta\chi^2 = 1$ tells the $\pm 1\sigma$ uncertainty range for the parameter



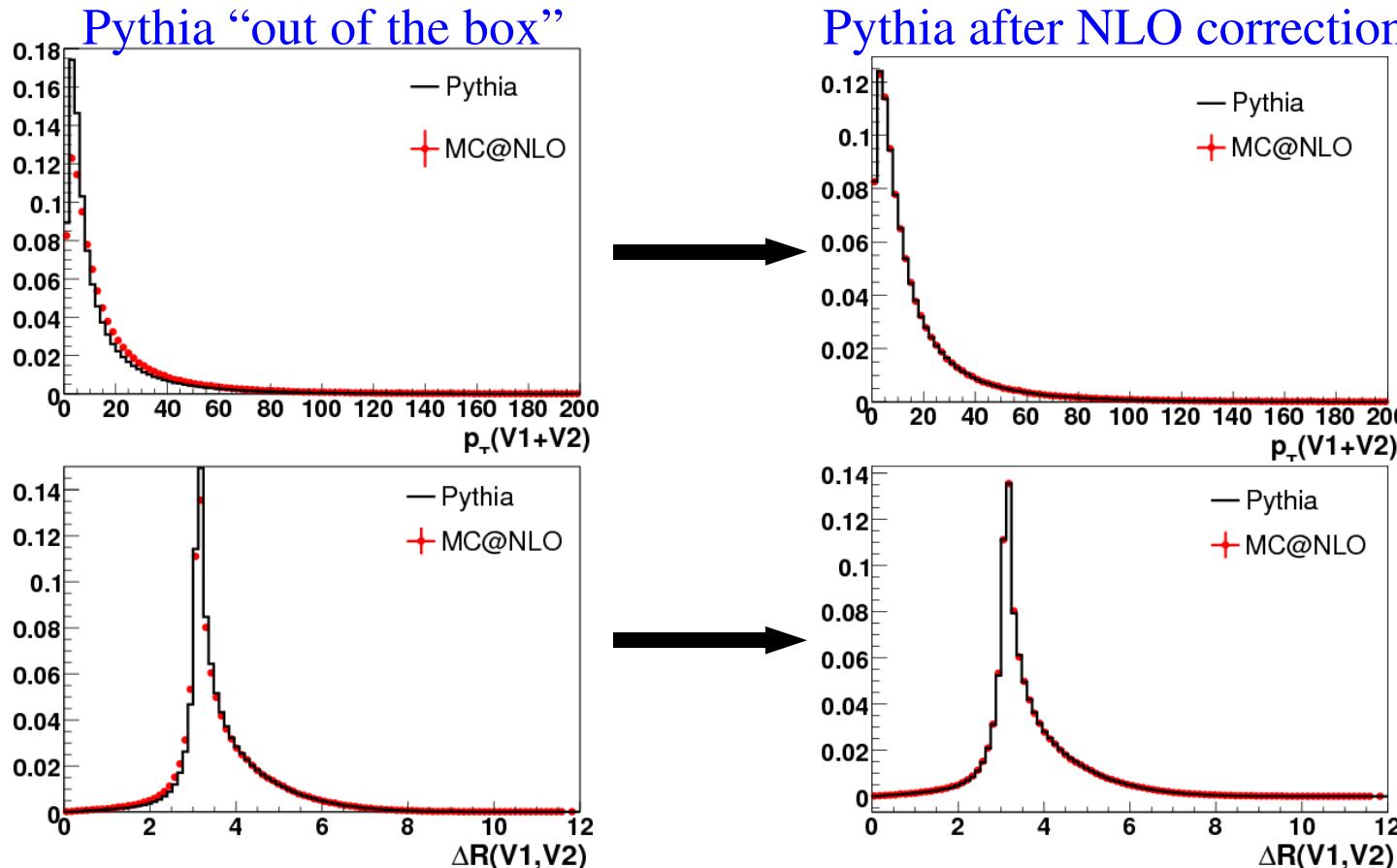
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 - Minimum χ^2 tells the parameter value that best models data
 - $\Delta\chi^2 = 1$ tells the $\pm 1\sigma$ uncertainty range for the parameter
- Data modeled best with **$E_T^{\text{Clus}} = 13.2 \text{ GeV}$**
 - ALPGEN authors recommend 13 GeV



Diboson $\mathcal{N}\text{LO}$ Correction

- Diboson signal generated at leading order
 - Use MC@NLO generator to correct signal MC to next-to-leading order
 - Generator level using a two dimensional re-weighting





Other Corrections

- Efficiency, reconstruction, resolution differences between data and MC
 - Trigger Efficiencies
 - Data must pass the trigger to be selected \Rightarrow apply these efficiencies to the MC
 - Jet (Lepton) Energy Scale, Energy Resolution, and Identification
 - MC does not do a perfect job of modeling detector response to jets (and to a much lesser extent, leptons) \Rightarrow correct energies and apply data/MC scale factors
 - Instantaneous Luminosity Profile
 - MC overlayed with zero-bias data to account for other activity in the detector
 - Luminosity profile for zero-bias not the same as for the data \Rightarrow re-weight MC events to have the same profile
 - z Position of Primary Vertex
 - Slightly non-Gaussian in data \Rightarrow re-weight MC to match
 - Muon Track Efficiency
 - Poor MC modeling of muon track with low curvature significance
 \Rightarrow remove these events
- \Rightarrow In total, 28 independent systematic uncertainties



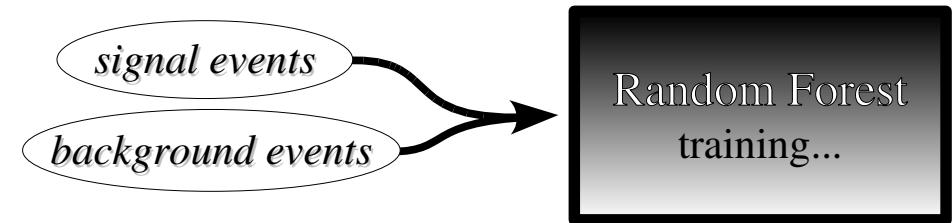


Multivariate Discrimination



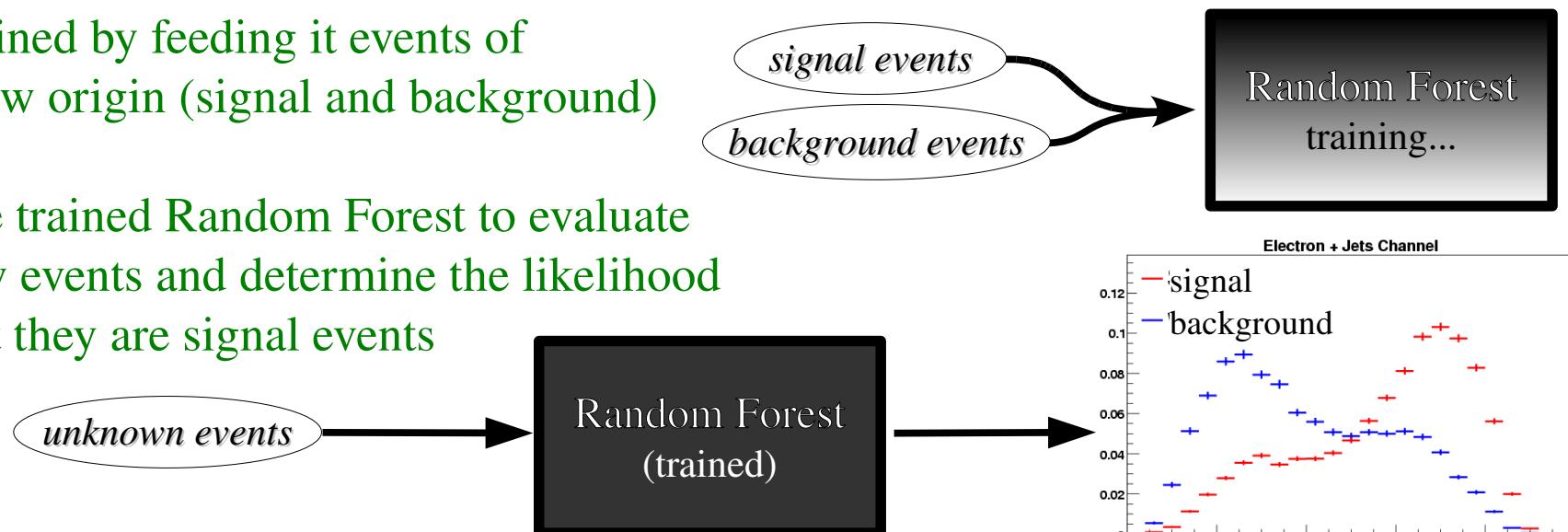
Multivariate Discrimination

- Use a multivariate classifier to improve separation of signal and background
 - Many options: Neural Networks, Decision Trees, Matrix Elements Method, etc.
 - Random Forest classifier powerful and robust
 - Similar technology as boosted decision trees used in single-top analyses
- From the outside (black box), the Random Forest works similar to other multivariate classifiers
 - Trained by feeding it events of known origin (signal and background)



Multivariate Discrimination

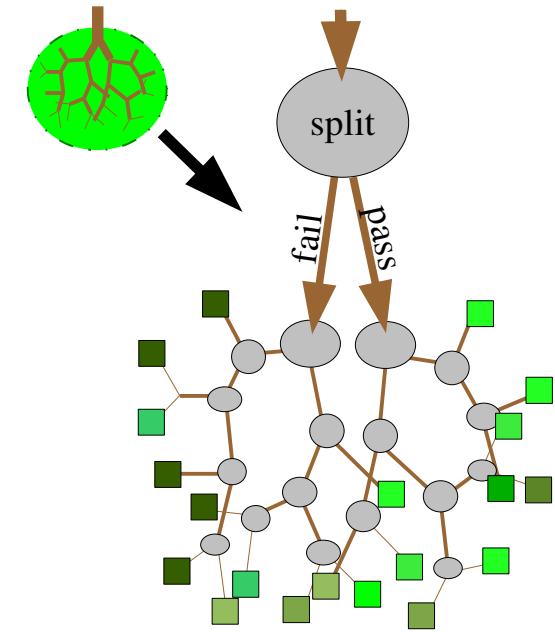
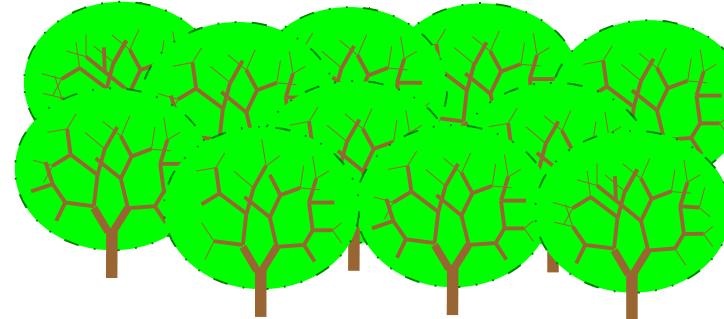
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 - Similar technology as boosted decision trees used in single-top analyses
- From the outside (black box), the Random Forest works similar to other multivariate classifiers
 - Trained by feeding it events of known origin (signal and background)
 - Use trained Random Forest to evaluate new events and determine the likelihood that they are signal events



Inside the Random Forest



A “forest” of many decision tree classifiers

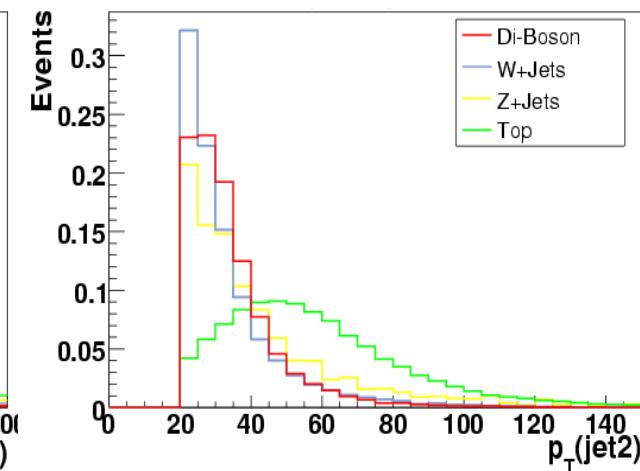
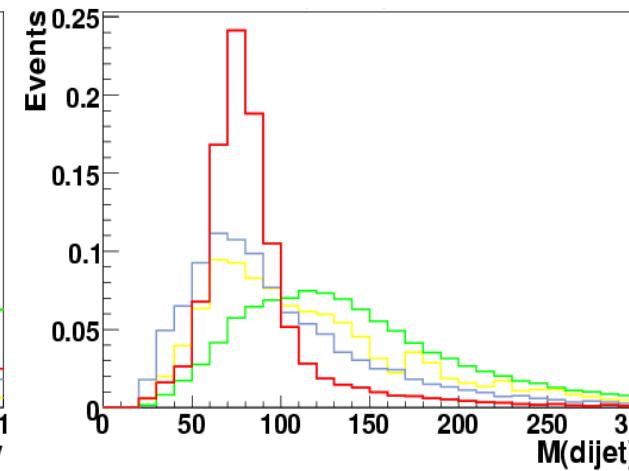
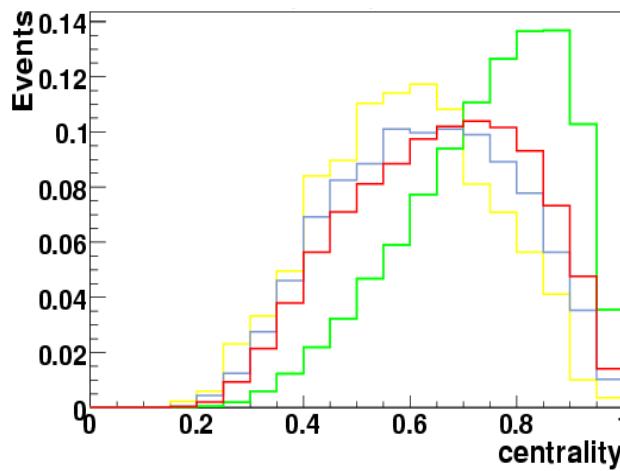


- Each tree is independent of the other trees
 - Each uses a random subset of input variables
 - ▶ Allows each tree to focus on a different subset of kinematics and correlations
 - Each is trained with a random subset of training events
 - ▶ Provides protection against over-training and high-weight events
- The Random Forest output averages the output from all the trees
 - Fluctuations and over-training that occur for a single decision tree are reduced because each tree will fluctuate differently

Multivariate Discrimination

- Choosing input variables to the Random Forest
 - Variables that are well modeled by the MC
 - χ^2 probability between data and MC greater than 5%
 - Variables that differ between signal and at least one background
 - However, well-modeled uninformative variables will not degrade performance

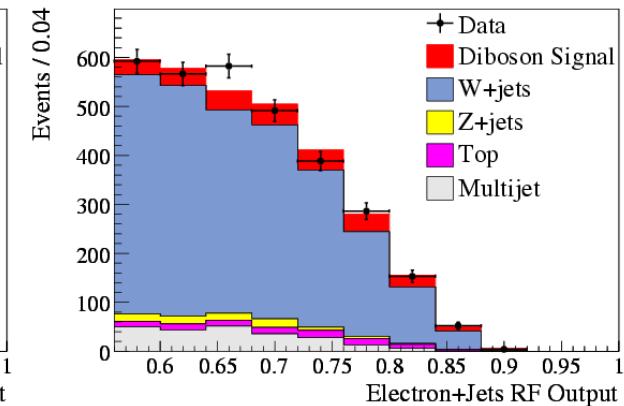
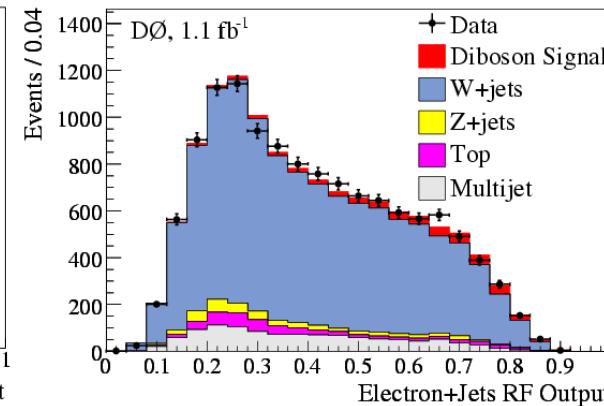
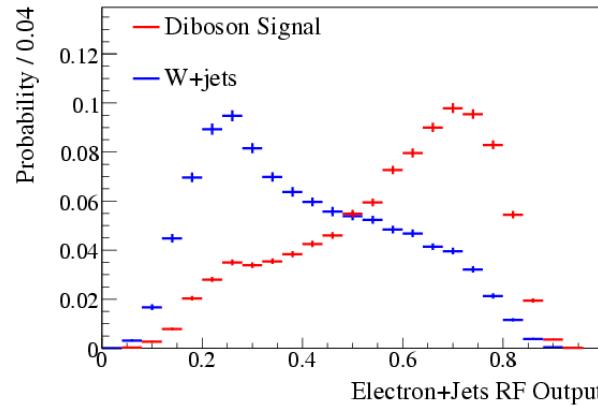
⇒ Came up with 13 input variables



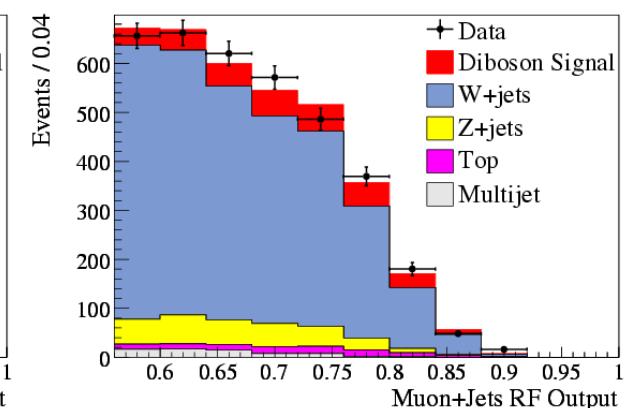
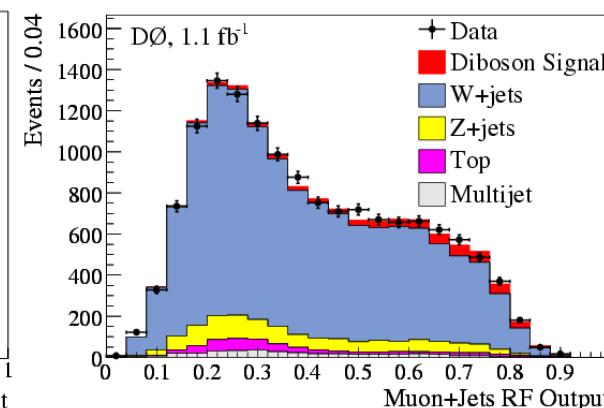
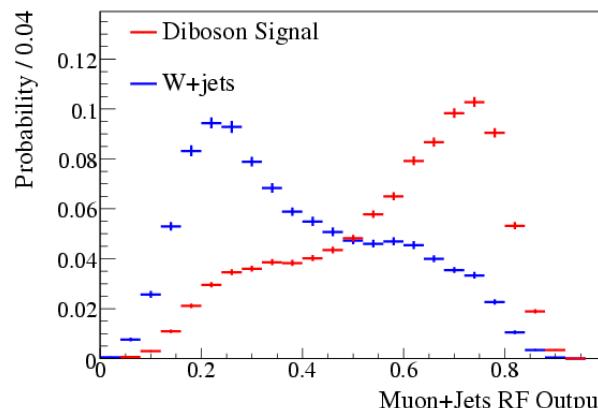
Multivariate Discrimination

- Output from the Random Forest
 - Discriminating variable with improved signal and background separation

$e\nu qq$ channel



$\mu\nu qq$ channel





The Measurement





Cross Section Measurement

- Perform “best fit” of distributions for signal and each background to the data
 - Predicted number of events, p_i , varied within the systematic uncertainties $\sigma_{k,i}$

$$p'_i = p_i \prod_k^{N_{\text{sys}}} (1 + R_k \sigma_{k,i})$$

- Minimizes the negative-log-likelihood between data and prediction

$$\text{NLL} = -2 \ln \mathcal{L} = 2 \sum_i^{N_{\text{bins}}} \left[p'_i - d_i - d_i \ln \left(\frac{p'_i}{d_i} \right) \right] + \sum_k^{N_{\text{sys}}} (R_k)^2$$





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Ratio of Poisson likelihoods

↑ Gaussian prior
constraint for systematics





Cross Section Measurement

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Ratio of Poisson likelihoods Gaussian prior constraint for systematics

- Systematics “floated” by removing the $(R_k)^2$ term \Rightarrow free fit parameters
 - Float W+jets normalization \Rightarrow determined by sideband regions
 - Float signal cross section \Rightarrow fit the excess (over background) that is consistent with kinematics of WW and WZ production

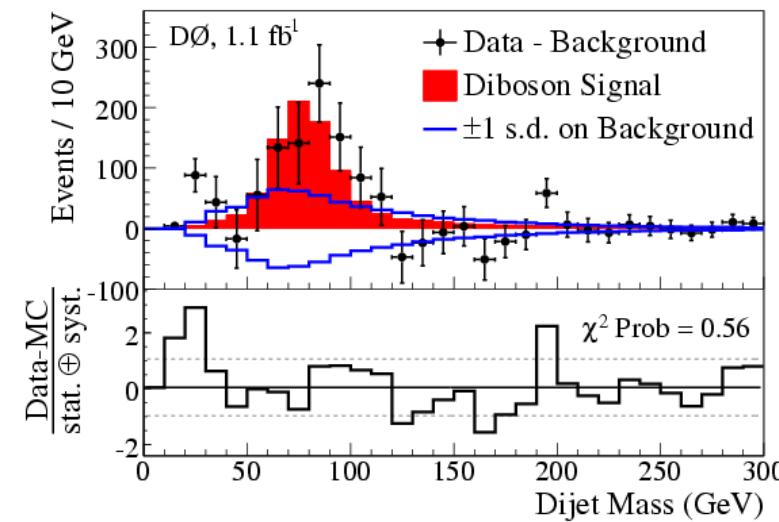
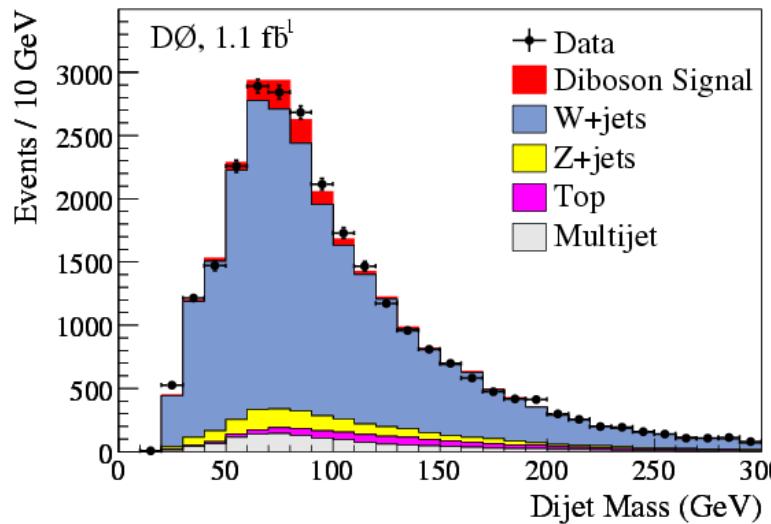


Cross Section Measurement

- Measured cross section using the **dijet mass**
⇒ $\sigma(WW+WZ) = 18.5 \pm 2.8(\text{stat}) \pm 4.9(\text{syst}) \pm 1.1(\text{lumi}) \text{ pb}$

(SM prediction: $\sigma(WW+WZ) = 16.1 \pm 0.9 \text{ pb}$)

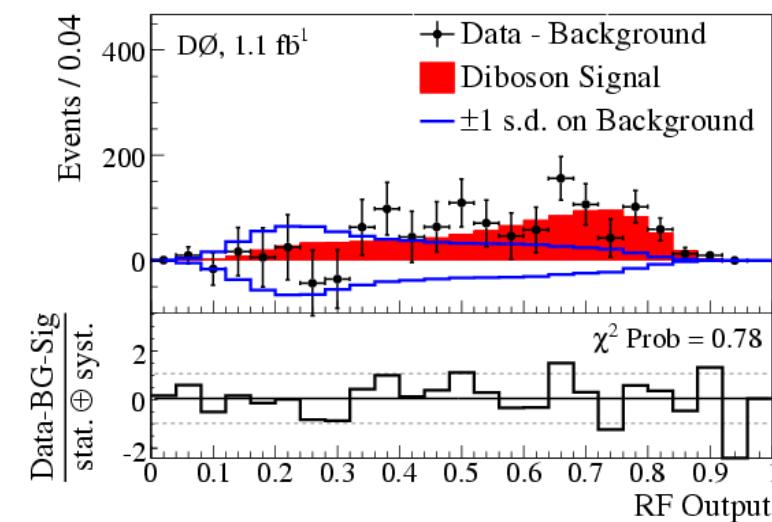
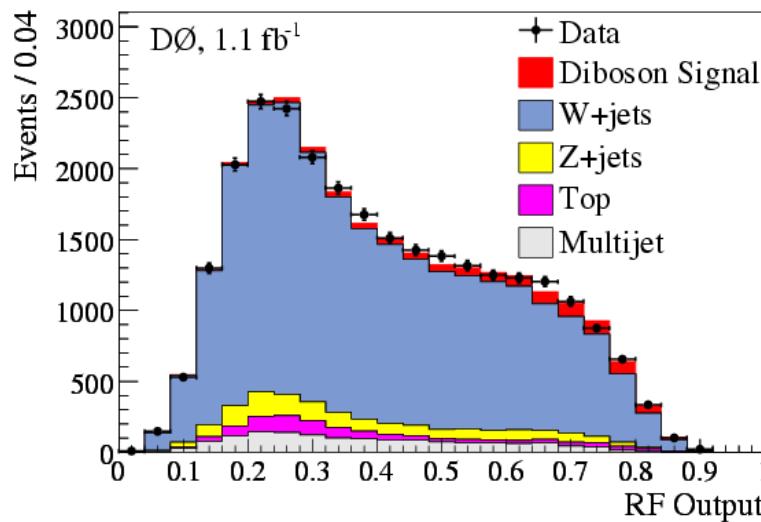
Dijet mass after best-fit using the dijet mass



Cross Section Measurement

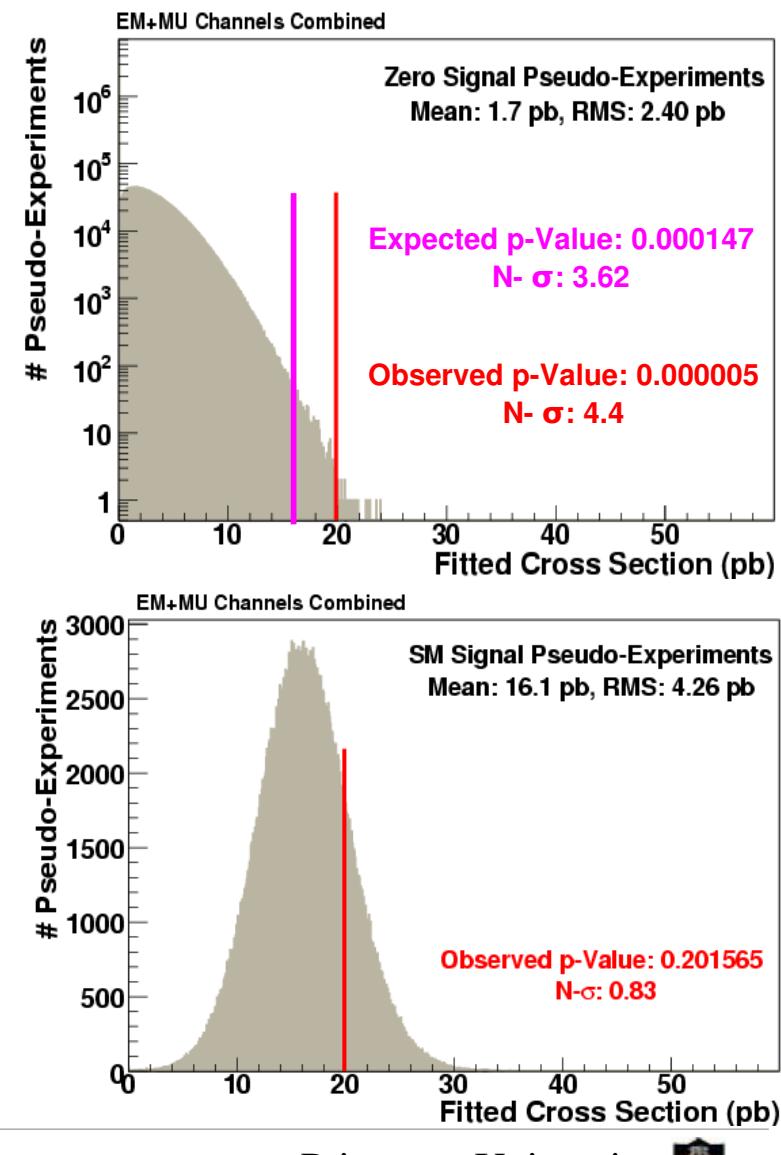
- Measured cross section using the **dijet mass**
⇒ $\sigma(WW+WZ) = 18.5 \pm 2.8(\text{stat}) \pm 4.9(\text{syst}) \pm 1.1(\text{lumi}) \text{ pb}$
- Measured cross section using the **Random Forest**
⇒ $\sigma(WW+WZ) = 20.2 \pm 2.5(\text{stat}) \pm 3.6(\text{syst}) \pm 1.2(\text{lumi}) \text{ pb}$
(SM prediction: $\sigma(WW+WZ) = 16.1 \pm 0.9 \text{ pb}$)

Random Forest after best-fit using the Random Forest



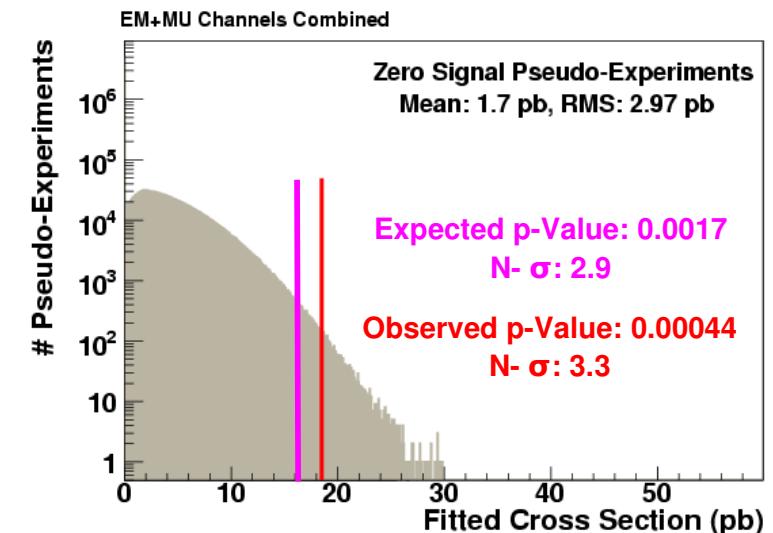
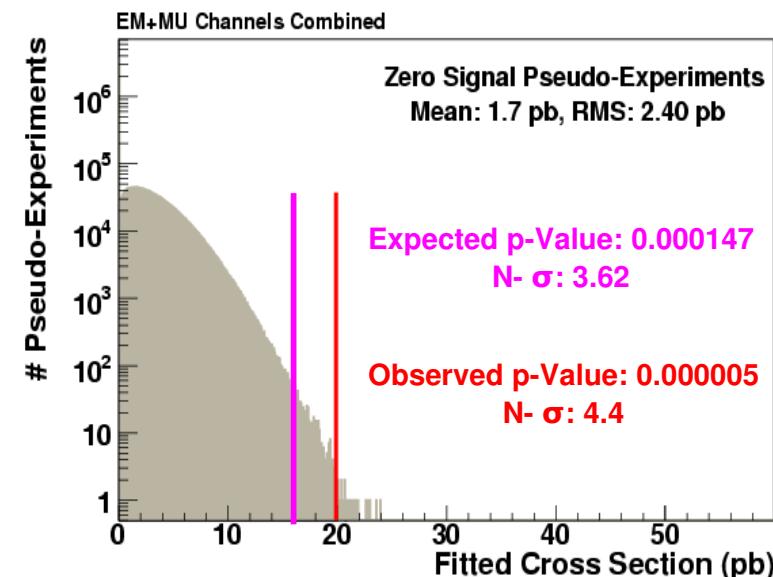
Significance

- Estimation of significance
- Frequentist approach: count fraction of outcomes for which the background-only scenario yields a measured cross section as large as we observe (or expect)
 - Generate pseudo-data for background-only scenario
 - randomly sample the systematics from their priors, then drawing Poisson trials for each bin
 - Fit the background-only pseudo-data outcomes in the same was as real data
- ⇒ Expected significance: 3.6σ (p-value = $1.5 \cdot 10^{-4}$)
- ⇒ Observed significance: 4.4σ (p-value = $5.4 \cdot 10^{-6}$)
- Compatibility with the SM
- Same approach as calculating the significance, except with pseudo-data for the signal+background scenario
 - ⇒ 0.83σ above SM prediction



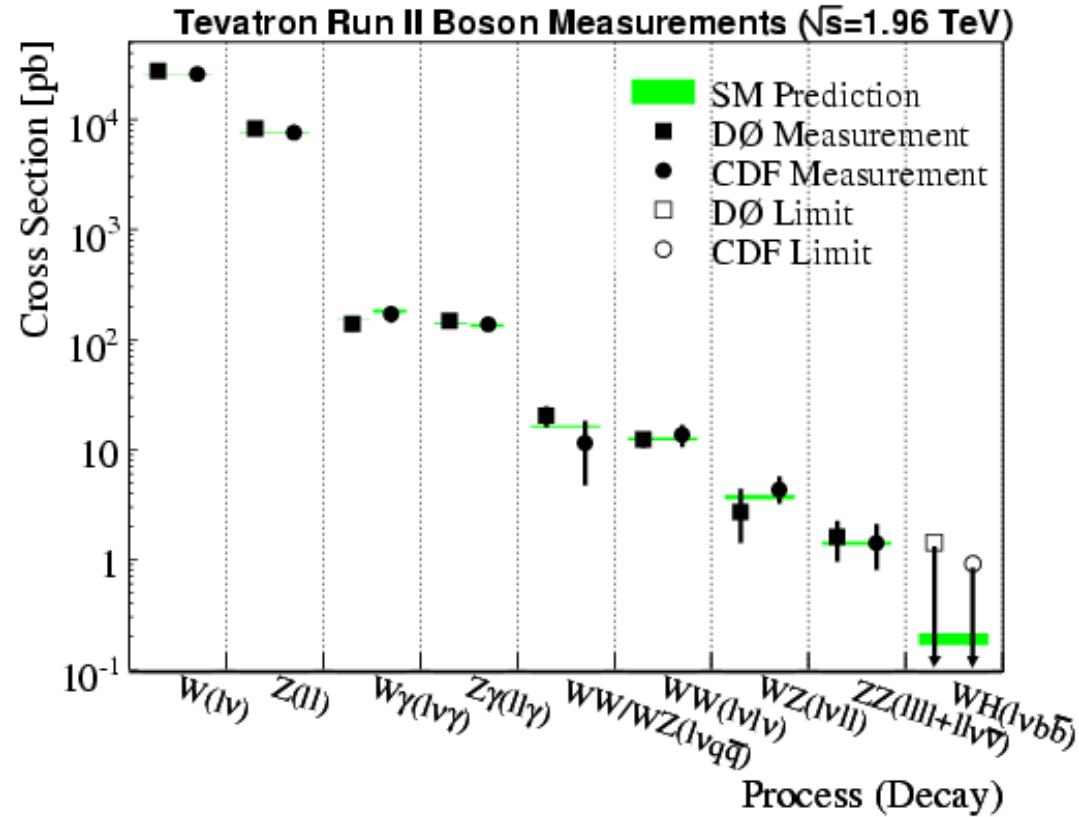
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- Significance using only dijet mass:
 - ⇒ Expected significance: 2.9σ (p-value = $1.7 \cdot 10^{-3}$)
 - ⇒ Observed significance: 3.3σ (p-value = $4.4 \cdot 10^{-4}$)



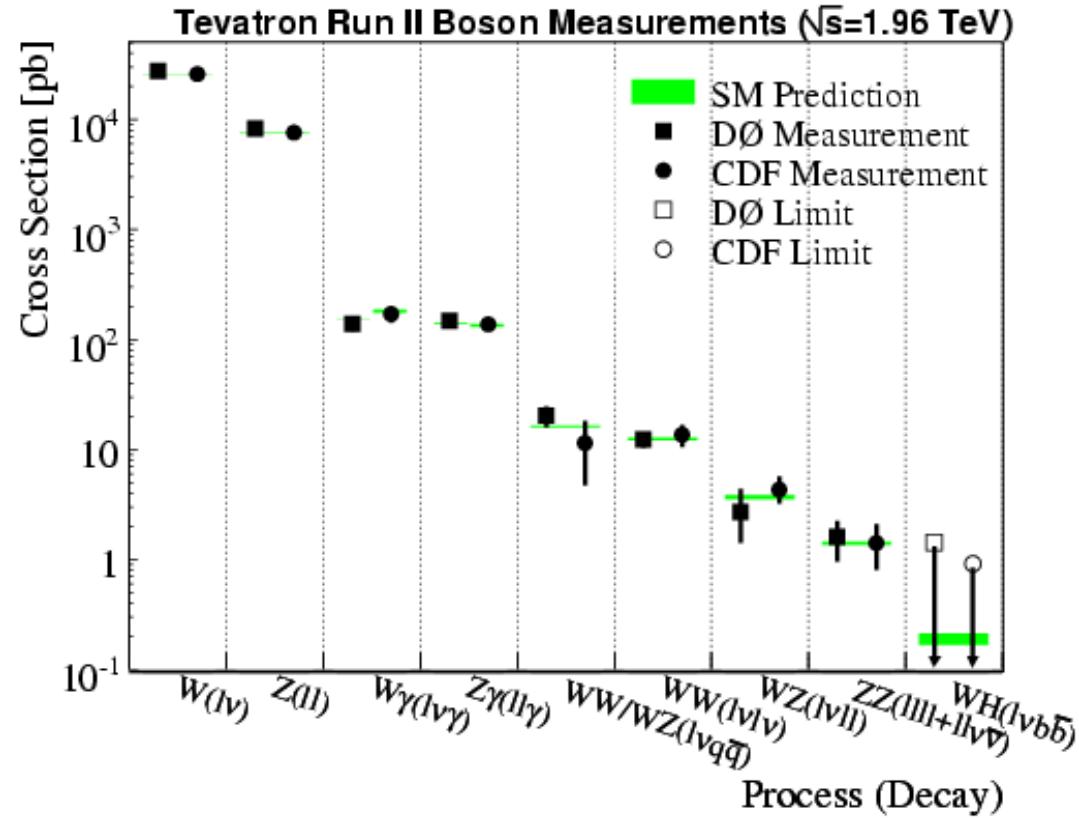
Summary

- First semi-leptonic diboson measurement at hadron collider
 - $\sigma(p\bar{p} \rightarrow WW + WZ) = 20.2 \pm 4.5 \text{ pb}$
 - Observed significance: 4.4σ
 - Accepted by PRL 3/25/09
- Only benchmark for low mass Tevatron Higgs analyses
 - First hadronic W mass outside of top quark events
- Direct constraint on dominant $H \rightarrow WW$ background
- Complimentary to and consistent with fully leptonic measurements
 - And more precise than previous leptonic measurements



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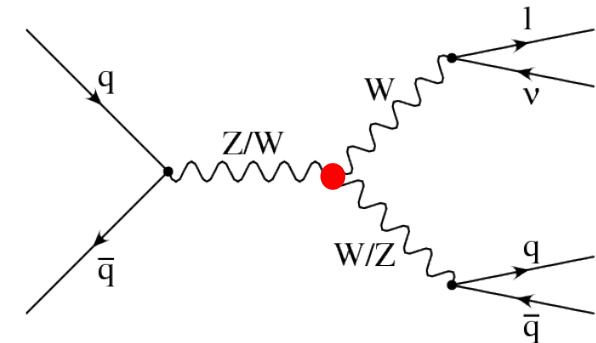


...what's next?



Continuing Work

- Limits on Anomalous Couplings
 - Compare the dijet p_T (most sensitive variable) between data and the predictions for different anomalous couplings
 - Results soon!
- Updating measurement to full Run II dataset ($>4 \text{ fb}^{-1}$)
 - Should achieve $>5\sigma$ significance \Rightarrow observation
 - Better (best Tevatron) limits on anomalous couplings
- Measure $Z \rightarrow bb$ Resonance
 - Apply b-tagging and measure contribution from only $WZ \rightarrow lvbb$
 - Also need to add $ZZ \rightarrow llbb$ channels
 - Expect sensitivity around 3σ
 - Analysis identical to $WH \rightarrow lvbb$ and $ZH \rightarrow llbb$, but $4 \times$ more signal
 - Would be the first evidence of a bb resonance at the Tevatron





thank you





Backup Slides

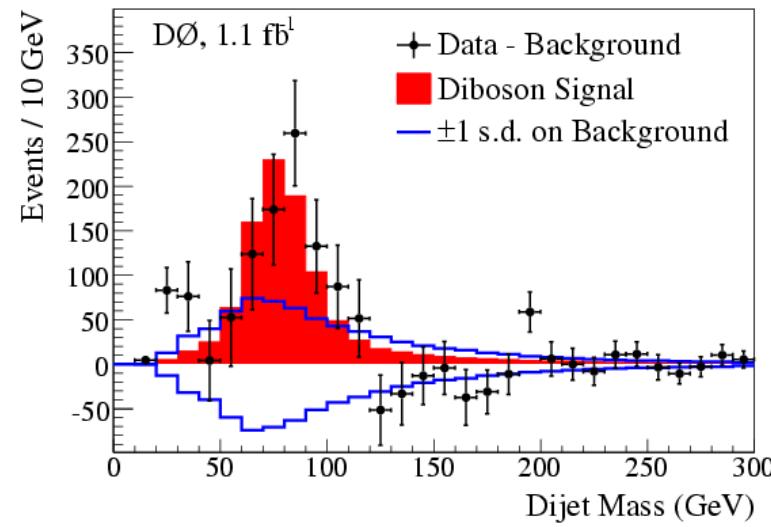


Comparison of Measurements

- Measuring $WW+WZ$ cross section
 - ◆ Using the **dijet mass**:
 $\sigma(WW+WZ) = 18.5 \pm 2.8(\text{stat}) \pm 4.9(\text{syst}) \pm 1.1(\text{lumi}) \text{ pb}$
 - ◆ Using the **random forest**:
 $\sigma(WW+WZ) = 20.2 \pm 2.5(\text{stat}) \pm 3.6(\text{syst}) \pm 1.2(\text{lumi}) \text{ pb}$

- Significance
 - ◆ Using the **dijet mass**:
Expected = 2.9σ
Observed = 3.3σ
 - ◆ Using the **random forest**:
Expected = 3.6σ
Observed = 4.4σ

Dijet mass distribution
after best-fit of Random Forest





Summary of Systematics

Percent change from nominal distribution for Random Forest distribution.

Source of systematic uncertainty	Diboson	$W+\text{jets}$	$Z+\text{jets}$	Top	Multijet	Nature
Trigger efficiency, $e\nu q\bar{q}$	+2/-3	+2/-3	+2/-3	+2/-3		N
Trigger efficiency, $\mu\nu q\bar{q}$	+0/-5	+0/-5	+0/-5	+0/-5		S
Lepton identification	± 4	± 4	± 4	± 4		N
Jet identification	± 1	± 1	± 1	$\pm <1$		S
Jet energy scale	± 4	± 9	± 9	± 4		S
Jet energy resolution	± 3	± 4	± 4	± 4		N
Cross section		$\pm 20^a$	± 6	± 10		N
Heavy flavor jet contribution		± 20				N
Diboson NLO correction	± 10					S
PDF set	± 1	± 1	± 1	± 1		S
ALPGEN jet η and ΔR corrections		± 1	± 1			S
Renormalization and factorization scale		± 3	± 3			S
ALPGEN parton-jet matching parameters		± 4	± 4			S
Multijet normalization, $e\nu q\bar{q}$					± 20	N
Multijet normalization, $\mu\nu q\bar{q}$					± 30	N
Multijet shape, $e\nu q\bar{q}$					± 6	S
Multijet shape, $\mu\nu q\bar{q}$					± 10	S





Summary of Systematics

Contribution of each systematic to the total systematic uncertainty of 3.6 pb on measured cross section using the Random Forest classifier.

Source of systematic uncertainty	$\Delta\sigma$ [pb]
Trigger efficiency, $e\nu q\bar{q}$	< 0.1
Trigger efficiency, $\mu\nu q\bar{q}$	< 0.1
Lepton identification	< 0.1
Jet identification	0.3
→ Jet energy scale	1.9
Jet energy resolution	< 0.1
Cross section	1.1
Heavy flavor jet contribution	< 0.1
Diboson NLO correction	< 0.1
PDF set	0.2
ALPGEN jet η and ΔR corrections	< 0.1
Renormalization and factorization scale	0.9
→ ALPGEN parton-jet matching parameters	2.4
Multijet normalization, $e\nu q\bar{q}$	0.9
Multijet normalization, $\mu\nu q\bar{q}$	0.5
Multijet shape, $e\nu q\bar{q}$	< 0.1
Multijet shape, $\mu\nu q\bar{q}$	< 0.1



Making the Measurement

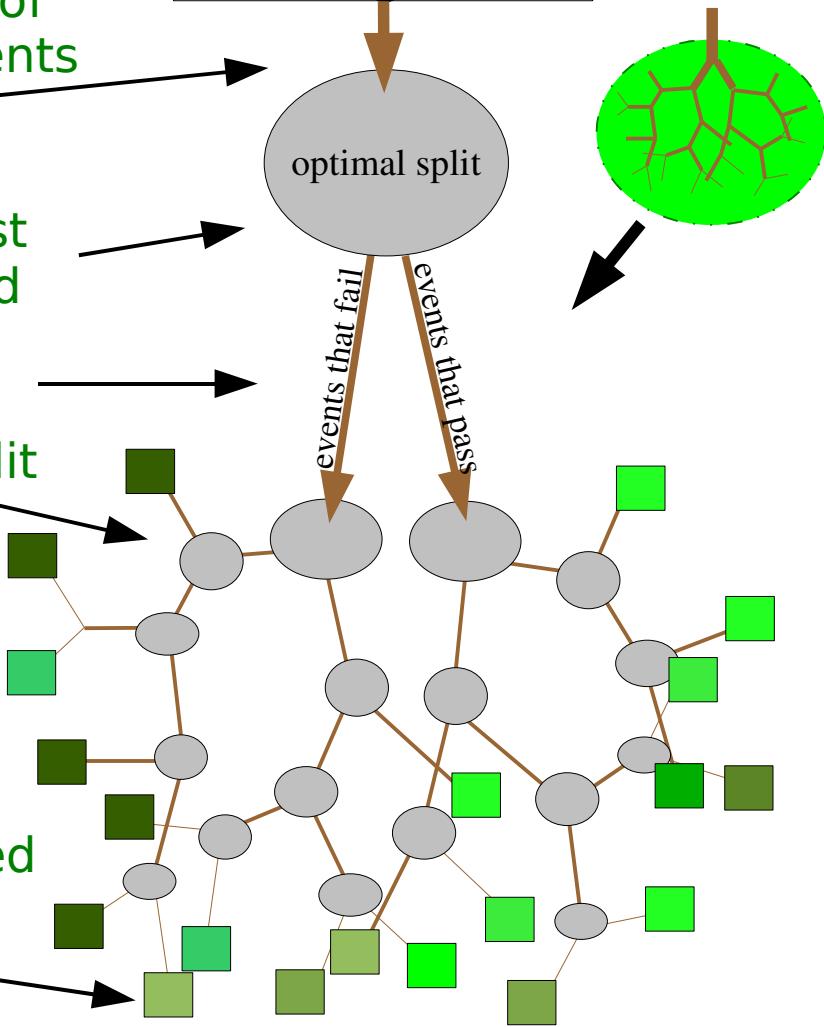
- Now that an adequate MC description of the data has been established
 - The random forest classifier should provide improved separation of signal events
 - Improved signal significance compared to individual variables, such as dijet mass
- The next step is to extract the signal
 - Cross section
 - Measure excess over background that is consistent with diboson production
 - Significance
 - Determine the probability the measured signal is really just a fluctuation in the background



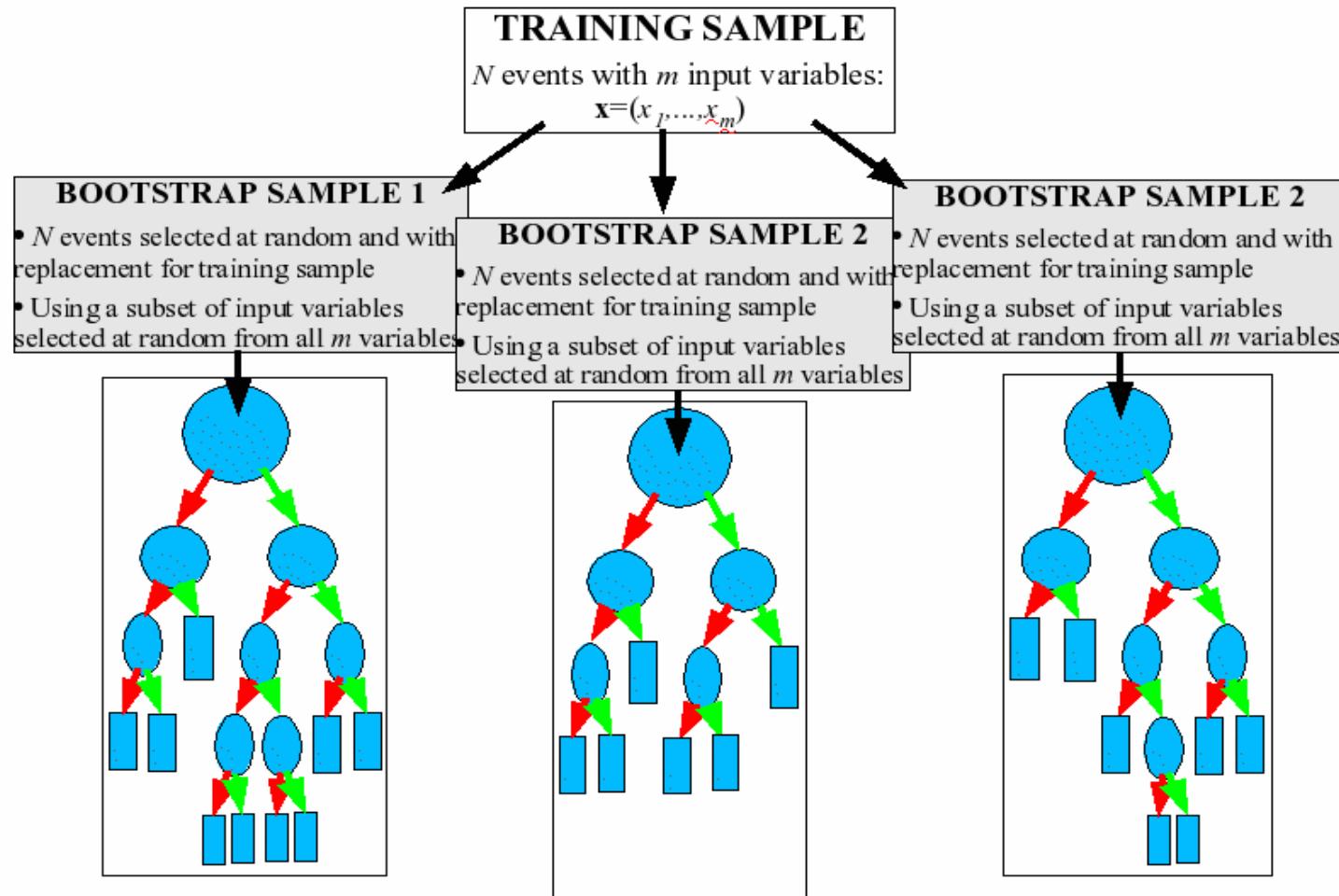
Decision Tree Classifier

- Decision tree is trained/grown using a set of known signal and background training events
⇒ These events go into the root node
- Algorithm looks at all possible splits on all input variables and applies split giving best separation between signal and background (using Gini index)
- Events pass to one of two child nodes depending on whether they pass or fail split
- This process is repeated for each child node until
 - a node contains all or no signal events
 - the number of events per node is less than a some specified amount (must be optimized for each analysis)
- Output for an unknown event is determined by the signal purity of the terminal node that the event ends up in

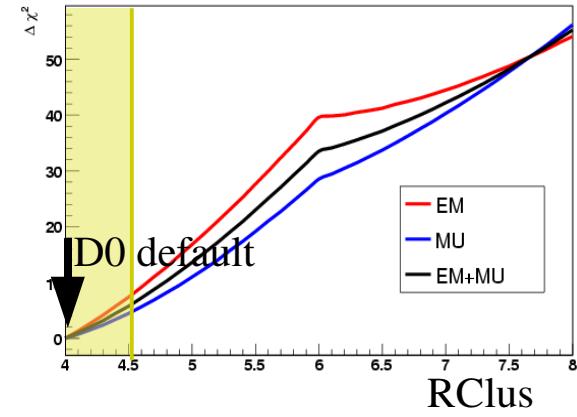
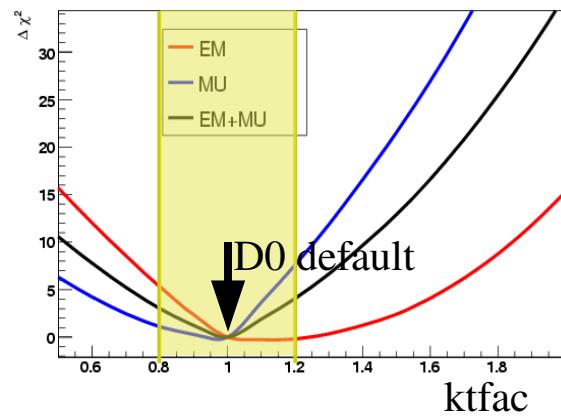
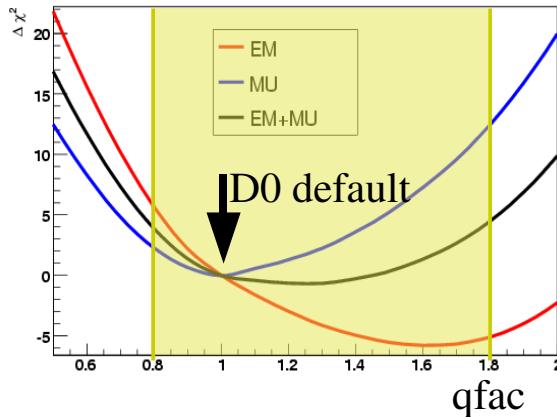
Training events



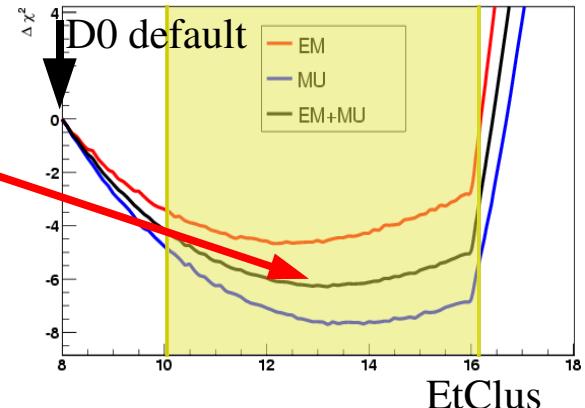
Random Forest Classifier



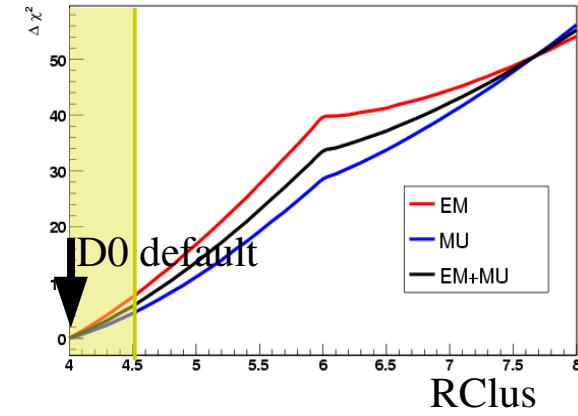
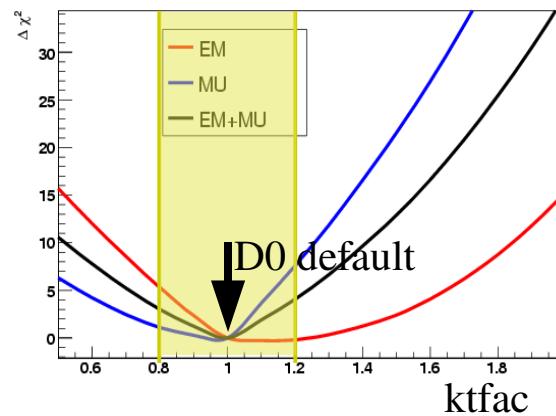
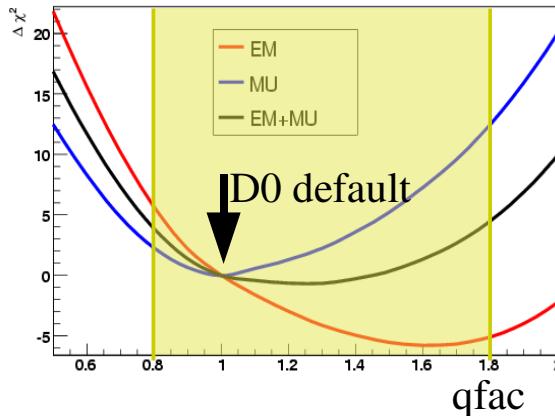
ALPGEN Modeling: Generator Parameters



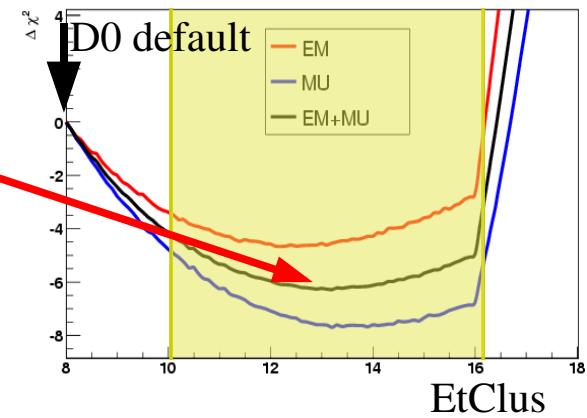
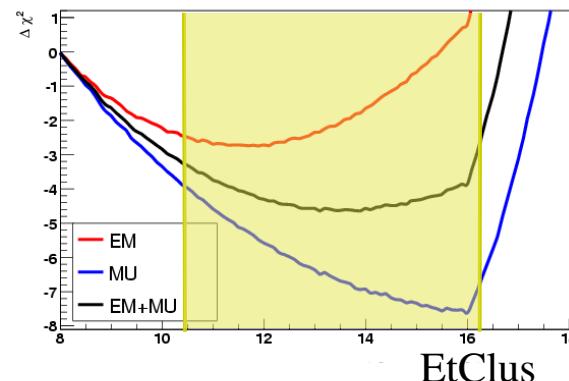
- Best values for $qfac$, $ktfac$, and $RClus$ are consistent with default values used by D0
- $EtClus$ models the data best for a value of 13.2 GeV
 - Value recommended by Alpgen authors is 13 GeV
 - ⇒ Re-weight Alpgen V+jets MC to $EtClus = 13.2$ GeV



ALPGEN Modeling: Generator Parameters



- Best values for qfac, ktfac, and RClus are consistent with default values used by D0
- EtClus models the data best for a value of 13.2 GeV
 - Value recommended by Alpgen authors is 13 GeV
⇒ Re-weight Alpgen V+jets MC to EtClus = 13.2 GeV
- Dijet mass shows similar behavior for EtClus (and other parameters) →



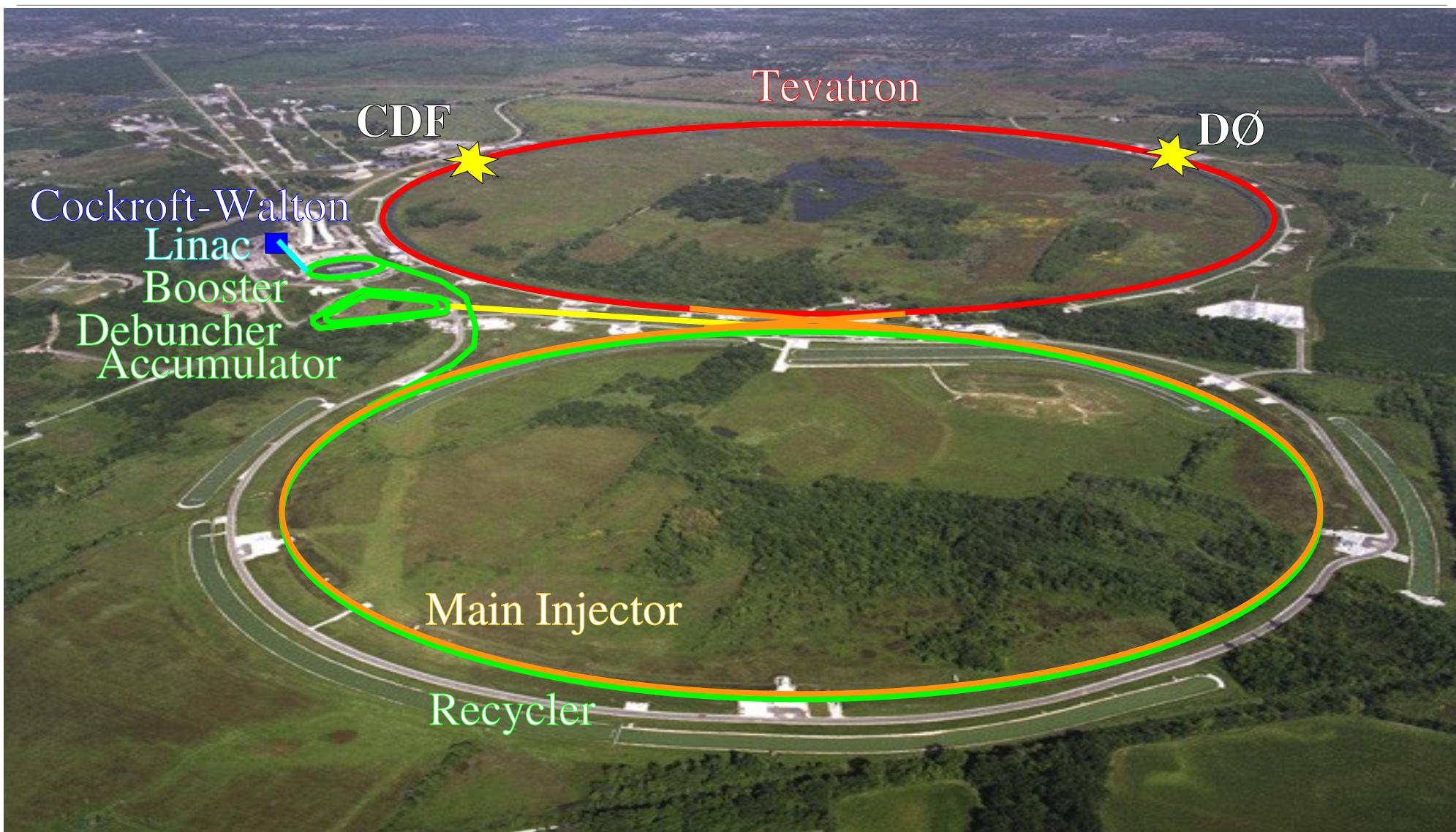


The *D*O Collaboration





The Accelerator Chain



Event Reconstruction

